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# The Hitchhiker's Guide to Logical Verification

2024 Desktop Edition

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lean-forward.github.io/ hitchhikers-guide/2024

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# **Preface**

Formal proof assistants are software tools designed to help their users carry out computer-checked proofs. We usually call them *proof assistants*, or *interactive theorem provers*, but a frustrated student coined the phrase "proof-preventing beasts," and dictation software occasionally misunderstands "theorem prover" as "fear improver." Consider yourself warned.

**Rigorous and Formal Proofs** Interactive theorem proving has its own terminology, starting with the notion of "proof." A *formal proof* is a logical argument expressed in a logical formalism. In this context, "formal" means "logical" or "logic-based." Logicians—the mathematicians of logics—carried out formal proofs on paper decades before the advent of computers, but nowadays formal proofs are almost always carried out using a proof assistant.

In contrast, an *informal proof* is what a mathematician would normally call a proof. These are often carried out in <u>MEX</u> or on a blackboard, and are also called "pen-and-paper proofs." The level of detail can vary a lot, and phrases such as "it is obvious that," "clearly," and "without loss of generality" move some of the proof burden onto the reader. A *rigorous proof* is a very detailed informal proof.

The main strength of proof assistants is that they help develop highly trustworthy, unambiguous proofs of mathematical statements, using a precise logic. They can be used to prove arbitrarily advanced results, going well beyond toy examples and logic puzzles. Formal proofs also help students understand what constitutes a valid definition or a valid proof. To quote Scott Aaronson:<sup>1</sup>

I still remember having to grade hundreds of exams where the students started out by assuming what had to be proved, or filled page after page with gibberish in the hope that, somewhere in the mess, they might accidentally have said something correct.

When we develop a new theory, formal proofs can help us explore it. They are useful when we generalize, refactor, or otherwise modify an existing proof, in much the same way as a compiler helps us develop correct programs. They offer a high level of trustworthiness, making it easier for others to review them. In addition, formal proofs can form the basis of verified computational tools (e.g., verified computer algebra systems).

**Success Stories** There have been a number of proof assistant success stories in mathematics and computer science. Some landmark results in the formalization of mathematics have been the proof of the four-color theorem by Gonthier et

https://www.scottaaronson.com/teaching.pdf

al. [6], the proof of the odd-order theorem by Gonthier et al. [7], the proof of the Kepler conjecture by Hales et al. [10], and the definition of perfectoid spaces by Buzzard et al. [4]. The earliest work in this area was carried out by Nicolaas de Bruijn and his colleagues starting in the 1960s in a system called AUTOMATH.<sup>2</sup>

Today, few mathematicians use proof assistants, but this is slowly changing. For example, 29 participants of the Lean Together 2019 meeting in Amsterdam,<sup>3</sup> about formalization of mathematics, self-identified as mathematicians.

Most users of proof assistants are computer scientists. A few companies, including AMD [27] and Intel [11], have been using proof assistants to verify their designs. In academia, some of the milestones are the verifications of the operating system kernels seL4 [14] and CertiKOS [9] and the development of the verified compilers CompCert [17], JinjaThreads [20], and CakeML [16].

**Proof Assistants** There are dozens of proof assistants in development or use across the world. A list of the main ones follows, grouped by their logical foundation:

- set theory: Isabelle/ZF, Metamath, Mizar;
- simple type theory: HOL4, HOL Light, Isabelle/HOL;
- dependent type theory: Agda, Coq, Lean, Matita, PVS;
- Lisp-like first-order logic: ACL2.

For a history of proof assistants and interactive theorem proving, we refer to Harrison, Urban, and Wiedijk's highly informative chapter [12].

**Lean** Lean is a proof assistant developed primarily by Leonardo de Moura (Microsoft Research) since 2012. Its mathematical library, mathlib, was originally developed under the leadership of Jeremy Avigad (Carnegie Mellon University) but is now maintained and further extended by the user community [21].<sup>4</sup>

This guide uses Lean version v4.6.0-rc1, mathlib revision 29c1f0ad4669e340, and a few extensions collected in a small library called LoVelib.<sup>5</sup> Although it is a research project, with some rough edges, there are several reasons why Lean is suitable to teach interactive theorem proving:

- It has a highly expressive, and very interesting, logic based on the calculus of inductive constructions, a dependent type theory.
- It is extended with classical axioms and quotient types, making it useful to verify mathematics.
- It includes a convenient metaprogramming framework, which can be used to program custom proof automation.
- It includes a modern user interface via a Visual Studio Code plugin.
- It has highly readable, fairly complete documentation.
- It is open source.

<sup>&</sup>lt;sup>2</sup>https://www.win.tue.nl/automath/

<sup>3</sup>https://lean-forward.github.io/lean-together/2019/index.html

<sup>4</sup>https://github.com/leanprover-community/mathlib4

 $<sup>^{5}</sup> https://github.com/blanchette/logical\_verification\_2024/raw/main/lean/LoVelib.lean$ 

Lean's core library includes only basic algebraic definitions. More definitions are found in mathlib. Despite its name, mathlib is more than a mathematical library; it provides a lot of basic automation on top of Lean's core library, as one would expect from a modern proof assistant.

**This Guide** This guide was originally designed as a companion to the MSc-level course Logical Verification (LoVe) taught at the Vrije Universiteit Amsterdam. Its primary aim is to teach interactive theorem proving. Lean is the means, not an end in itself. As such, this guide is not designed to be a comprehensive Lean tutorial—for this, we recommend *Theorem Proving in Lean 4* [13]. This guide is also no substitute for doing the exercises and homework. Theorem proving is not for spectators; it can only be learned by doing.

Specifically, the aim is that you

- learn the fundamental theory and techniques of interactive theorem proving;
- learn how to use logic as a precise language for modeling systems and stating properties about them;
- become familiar with some areas in which proof assistants are successfully applied, such as functional programming, the semantics of imperative programming languages, and mathematics;
- develop practical skills that can be applyed in larger projects (whether personal or for an MSc or PhD or in industry);
- feel able to move to another proof assistant and apply what you have learned;
- get to understand the domain well enough to start reading relevant scientific papers published at international conferences such as Certified Programs and Proofs (CPP) and Interactive Theorem Proving (ITP) or in journals such as the Journal of Automated Reasoning (JAR).

Equipped with a good knowledge of Lean, it should be easy to move to another proof assistant based on dependent type theory, such as Agda or Coq, or to a system based on simple type theory, such as HOL4 or Isabelle/HOL.

An important characteristic of this guide, which it shares with Knuth's *T<sub>E</sub>Xbook* [15], is that it does not always tell the truth. To simplify the exposition, some basic but false claims are made about Lean. Most of these statements are then rectified in later chapters. Like Knuth, we feel that "this technique of deliberate lying will actually make it easier for you to learn the ideas. Once you understand a simple but false rule, it will not be hard to supplement that rule with its exceptions."

The Lean files accompanying this guide can be found in a public repository. The files' naming scheme follows this guide's chapters; thus, LoVeo7\_Effectful-Programming\_Demo.lean is the main file associated with Chapter 7 ("Effectful Programming"), reviewed in class, LoVeo7\_EffectfulProgramming\_ExerciseSheet. lean is the exercise sheet, and LoVeo7\_EffectfulProgramming\_HomeworkSheet. lean is the homework sheet.

We owe a huge debt to the authors of *Theorem Proving in Lean 4* [13] and *Concrete Semantics: With Isabelle/HOL* [23], who have taught us Lean and programming language semantics. Many of their ideas appear in this guide.

<sup>&</sup>lt;sup>6</sup>https://github.com/blanchette/logical\_verification\_2024

We thank Robert Lewis and Assia Mahboubi for their useful comments on this guide's organization and focus. We thank Kiran Gopinathan and Ilya Sergey for sharing the anecdote reported in footnote 3 of Chapter 2 and letting us share it further. We thank Daniel Fabian for designing the first tablet-optimized version of this guide. We thank Paul Chisholm for authoring some of the comments in the associated Lean files. We thank Mark Summerfield for his many textual suggestions. Finally, we thank Chris Bailey, Kevin Buzzard, Paul Chisholm, Dominique Danco, Raufs Dunamalijevs, Wan Fokkink, Lina Gerlach, Robert Lewis, Pietro Monticone, Antonius Danny Reyes, Robert Schütz, Patrick Thomas, Huub Vromen, Floris Westerman, and Wijnand van Woerkom for reporting typos and some more serious errors they found in earlier editions of this guide. Should *you* find some mistakes in this text, please let us know.<sup>7</sup>

**Special Symbols** In this guide, we assume that you will be using Visual Studio Code and its Lean extension to edit .lean files. Visual Studio Code lets you enter Unicode symbols by entering backslash \ followed by an ASCII identifier. For example,  $\rightarrow$ ,  $\forall$ , or  $\in$  can be entered by typing \->, \fo, or \in and pressing the tab key or the space bar. We will freely use these notations. For reference, we provide a list of the main non-ASCII symbols that are used in this guide and, for each, one of its ASCII representations. By hovering over a symbol in Visual Studio Code while holding the control or command key pressed, you will see the different ways in which it can be entered.

$\neg$	\not	$\wedge$	\and	V	\or
$\rightarrow$	\->	$\leftrightarrow$	\<->	$\forall$	\fo
$\exists$	\ex	$\leq$	\<=	$\geq$	<b>\&gt;=</b>
$\neq$	\neq	$\approx$	\~~	$\times$	\x
0	\circ	Ø	\empty	$\cup$	\union
$\cap$	\intersect	$\in$	\in	1	\downleftharpoon
$\circ$	\bigcirc	$\leftarrow$	\<-	$\mapsto$	\mapsto
$\Rightarrow$	\=>	$\Longrightarrow$	\==>		]]/
	\]]	α	\a	β	<b>\</b> b
γ	\g	δ	\de	ε	\e
σ	\s	0	\0	1	\1
2	\2	3	\3	4	\4
5	\5	6	\6	7	\7
8	\8	9	\9		

<sup>&</sup>lt;sup>7</sup>jasmin.blanchette@ifi.lmu.de

# Part I

# **Basics**

# **Chapter 1**

# **Types and Terms**

We begin our journey by studying the basics of Lean, starting with terms (also called expressions) and their types.

Lean's logical foundation is a rich logic called the *calculus of inductive constructions*, which supports *dependent types*. Lean's logic is inspired by the  $\lambda$ -calculus and resembles typed functional programming languages such as Haskell, OCaml, and Standard ML. Even if you have not used any of these languages, you will recognize many of the concepts from modern languages (e.g., C++, Java, Python). In a first approximation:

If your background is in mathematics, you probably already know most of the key concepts underlying functional programming, sometimes under different names. For example, the Haskell program

```
fib 0 = 0
fib 1 = 1
fib n = fib (n - 2) + fib (n - 1)
```

closely matches the mathematical definition

$$fib(n) = \begin{cases} 0 & \text{if } n = 0 \\ 1 & \text{if } n = 1 \\ fib(n-2) + fib(n-1) & \text{if } n \ge 2 \end{cases}$$

In this chapter, we restrict our attention to a fragment of Lean without dependent types, called *simple type theory* (or *higher-order logic*). This corresponds roughly to the simply typed  $\lambda$ -calculus [3] extended with an equality operator (=). It is an abstract, extremely simplified version of a programming language with a function-calling mechanism that prefigures functional programming.

## 1.1 Types

Types can be basic types such as  $\mathbb{Z}$ ,  $\mathbb{Q}$ , and Bool or total functions  $\sigma \to \tau$ , where  $\sigma$  and  $\tau$  are themselves types. Types indicate which values an expression may evaluate to. They introduce a discipline that is followed somewhat implicitly in mathematics. In principle, nothing prevents a mathematician from stating  $1 \in 2$ , but a typing discipline would mark this as the error it likely is.

Semantically, types can be viewed as sets. We would normally define the types  $\mathbb{Z}$ ,  $\mathbb{Q}$ , and Bool so that they faithfully capture the mathematicians'  $\mathbb{Z}$  and  $\mathbb{Q}$  and the computer scientists' Booleans, and similarly for the function arrow ( $\rightarrow$ ). But despite their similarities, Lean and mathematics are distinct languages. Lean's types may be *interpreted* as sets, but they are not sets.

Higher-order types are types containing  $\to$  arrows that are nested on the left of  $\to$ , as in the type  $(\mathbb{Z} \to \mathbb{Z}) \to \mathbb{Q}$ . Values of such types are functions that take other functions as arguments. Accordingly,  $(\mathbb{Z} \to \mathbb{Z}) \to \mathbb{Q}$  is the type of unary functions that take a function of type  $\mathbb{Z} \to \mathbb{Z}$  as argument and that return a value of type  $\mathbb{Q}$ .

#### 1.2 Terms

The terms, or expressions, of simple type theory consist of

- constants c;
- variables x;
- applications t u;
- anonymous functions fun  $x \mapsto t$  (also called  $\lambda$ -expressions).

Above, t and u denote arbitrary terms. We can also write t:  $\sigma$  to indicate that the term t has the type  $\sigma$ .

Let us review each kind of term:

- A constant  $c:\sigma$  is a symbol of type  $\sigma$  whose meaning is fixed in the current global context. For example, an arithmetic theory might contain constants such as  $o:\mathbb{Z}$ ,  $1:\mathbb{Z}$ , abs  $:\mathbb{Z}\to\mathbb{N}$ , square  $:\mathbb{N}\to\mathbb{N}$ , and prime  $:\mathbb{N}\to Bool$ . Constants include functions (e.g., abs) and predicates (e.g., prime).
- A variable  $x : \sigma$  is either bound or free. A bound variable refers back to the input x of an anonymous function  $fun x : \sigma \mapsto t$  enclosing it. In contrast, a free variable is declared in the local context—a concept that will be explained below.
- An application t u, where t :  $\sigma \to \tau$  and u :  $\sigma$ , is a term of type  $\tau$  denoting the result of applying the function t to the argument u. For example, if t is abs and u is  $\sigma$ , then the application is abs  $\sigma$ . No parentheses are needed around the argument, unless it is a complex term—e.g., prime (abs  $\sigma$ ).
- Given a term  $t:\tau$ , an anonymous function fun  $x:\sigma\mapsto t$  denotes the total function of type  $\sigma\to\tau$  that maps each input value x of type  $\sigma$  to the body t, where x may occur in t. Thus, fun  $x:\mathbb{Z}\mapsto \text{square (abs }x)$  denotes the function that maps (the value denoted by)  $\circ$  to (the value denoted by) square (abs  $\circ$ ), that maps 1 to square (abs 1), and so on. The variable x is said to be bound by fun. Accordingly, fun is called a *binder*. For mathematicians, a more familiar syntax would have been  $x\mapsto \text{square (abs }x)$ , without the fun keyword.

Constants and variables look syntactically similar, but they are considered different. Constants are declared globally, whereas variables are introduced locally by fun or some other binder.

Applications and anonymous functions mirror each other: An anonymous function "constructs" a function; an application "destructs" a function. What happens if we combine the two? If we apply an argument u of type  $\sigma$  to an anonymous

1.2. Terms 5

function  $fun \ x : \sigma \mapsto t[x]$ , where t[x] represents some term that may contain x, then we obtain the term t[u]. (This is a slightly simplified view. Actually, if t[x] contains some binders, the bound variables might have to be renamed to avoid capturing u's free variables. In Lean, this renaming takes place automatically.) Here are a few examples of applying an anonymous function:

Although our functions are unary (i.e., they take one argument), we can build n-ary functions by nesting funs, using an ingenious technique called currying. For example, fun  $x: \sigma \mapsto (\text{fun } y: \tau \mapsto x)$  denotes the function of type  $\sigma \to (\tau \to \sigma)$  that takes two arguments and returns the first one. Strictly speaking,  $\sigma \to (\tau \to \sigma)$  takes a single argument and returns a function, which in turn takes an argument.

Applications work in the same way: If  $K := (fun \ x : \mathbb{Z} \mapsto (fun \ y : \mathbb{Z} \mapsto x))$ , then  $K = (fun \ y : \mathbb{Z} \mapsto 1)$  and (K = 1) = 1. The function K = 1, which is applied to a single argument, is said to be *partially applied*, because it can take an additional argument.

Currying is so useful a concept that we will omit most parentheses, writing

```
\begin{array}{ccc} \sigma \rightarrow \tau \rightarrow \upsilon & \text{for} & \sigma \rightarrow (\tau \rightarrow \upsilon) \\ & \text{tuv for } (\text{tu}) \ v \\ \text{fun } \text{x} : \sigma \mapsto \text{fun y} : \tau \mapsto \text{t} & \text{for } \text{fun x} : \sigma \mapsto (\text{fun y} : \tau \mapsto \text{t}) \end{array}
```

We will also use these abbreviations:

Moreover, in an anonymous function fun  $x:\sigma\mapsto t$ , we can usually omit the type annotation  $:\sigma$ , writing

```
fun x \mapsto t for fun x : \sigma \mapsto t
```

Lean will then try to infer the type based on the body t and on the context surrounding the anonymous function. Type inference lightens notations and saves some keystrokes.

In mathematics, it is customary to write binary operators in infix syntax (e.g., x + y). Such notations are also possible in Lean, as syntactic sugar for curried function application (e.g., add x y).

One way to work with Lean is to declare the types and constants we need using the opaque command. Consider the following declarations:

```
opaque a : \mathbb{Z}
opaque b : \mathbb{Z}
opaque f : \mathbb{Z} \to \mathbb{Z}
opaque g : \mathbb{Z} \to \mathbb{Z} \to \mathbb{Z}

#check fun x : \mathbb{Z} \mapsto g (f (g a x)) (g x b)
#check fun x \mapsto g (f (g a x)) (g x b)
```

The first four lines declare four constants (a, b, f, g), which can be used to form terms. The last two lines use the #check command to type-check terms and show their types. Using traditional mathematical notations, the term on the last line would be written  $x \mapsto g(f(g(a,x)),g(x,b))$ . The # prefix identifies diagnostic commands: commands that are useful for debugging but that we would not normally keep in our Lean files.

## 1.3 Type Checking and Type Inference

When Lean parses a term, it checks whether the term is well typed. In the process, it tries to infer the types of bound variables if those are omitted—e.g., the type of x in fun  $x \mapsto \text{square } (\text{square } x)$ .

For simple type theory, type checking and type inference are decidable problems. Advanced features such as overloading (the possibility to reuse the same name for several constants—e.g.,  $o:\mathbb{N}$  and  $o:\mathbb{R}$ ) can lead to undecidability [25]. Lean takes a pragmatic, computer-science-oriented approach and assumes that numerals  $o, 1, 2, \ldots$  are of type  $\mathbb{N}$  if several types are possible.

Lean's type system can be expressed as a formal system. A formal system consists of judgments and of derivation rules for producing judgments. A typing judgment has the form  $C \vdash t : \sigma$  and means that term t has type  $\sigma$  in local context C. For example, the judgment  $\vdash$  abs  $: \mathbb{Z} \to \mathbb{N}$  states that the constant abs has type  $\mathbb{Z} \to \mathbb{N}$  in the empty local context.

The local context gives the types of the variables in t that are not bound by a fun. It is used to keep track of the variables bound by binders *outside* t. If the same variable x is bound multiple times, the last occurrence shadows the other ones. For example, the judgment  $x : \mathbb{Z}$ ,  $y : \mathbb{N}$ ,  $y : \mathbb{Z} \vdash y : \mathbb{Z}$  states that the variable y has type  $\mathbb{Z}$  in the local context  $x : \mathbb{Z}$ ,  $y : \mathbb{N}$ ,  $y : \mathbb{Z}$ .

For simple type theory, the typing judgments are produced by four typing rules, one per kind of term:

Each rule has zero or more premises (above the horizontal bar), a conclusion (below the bar), and possibly a side condition (on the right). The premises are typing judgments, whereas the side conditions are arbitrary mathematical conditions on the mathematical variables occurring in the rule. To show the premises, we need to continue performing a derivation upward, as we will see in a moment. As for the side conditions, we can use the entire arsenal of mathematics to show that they are true.

The first two rules, labeled CST and VAR, have no premises, but they have side conditions that must be satisfied for the rules to apply. The last two rules take one

or two judgments as premises and produce a new judgment. Fun is the only rule that modifies the local context: As we enter the body t of an anonymous function, we need to record the existence of the bound variable x and its type to be ready when we meet x in t.

We can use this rule system to prove that a given term is well typed by working our way backwards (i.e., upwards) and applying the rules, building a *formal derivation* of a typing judgment. Like natural trees, derivation trees are drawn with the root at the bottom. The derived judgment appears at the root, and each branch ends with the application of a premise-less rule. Rule applications are indicated by a horizontal bar and a label. The following typing derivation establishes that the term fun  $x: \mathbb{Z} \mapsto \mathsf{abs}\ x$  has type  $\mathbb{Z} \to \mathbb{N}$  in an arbitrary local context C:

Reading the proof from the root upwards, notice how the local context is threaded through and how it is extended by the FUN rule. The rule moves the variable bound by fun to the local context, making an application of VAR possible further up the tree. If the variable x is already declared in C, it becomes shadowed by  $x:\mathbb{Z}$  after entering the fun expression.

In summary, the type system consists of derivation rules that can be (1) instantiated with arbitrary values for their mathematical variables and (2) connected to form derivation trees.

Here is a second example, this time starting with an empty local context:

Both examples so far were well typed. If we start with an ill-typed term, or if we specify the wrong type or context in the judgment at the root of the derivation tree, we will find that we cannot complete the derivation. A derivation constitutes a proof that a term is well typed and has the specified type in the given context.

The above type system checks only that terms are well typed. It does not check that types are well formed. For example, given the unary type constructor List, List  $\mathbb{Z}$  (the type of lists of integers) is well formed, whereas  $\mathbb{Z}$  List and List List are ill-formed. For simple type theory, well-formedness is easy to check: Only declared type constructors should be used, and each n-ary type constructor should be passed exactly n type arguments.

## 1.4 Type Inhabitation

Given a type  $\sigma$ , the type inhabitation problem consists of finding an "inhabitant" of that type—a term of type  $\sigma$ —within the empty local context. It may seem like a pointless exercise, but as we will see in Chapter 4, this problem is closely related

to that of finding a proof of a proposition. Seemingly silly exercises of the form "find a term of type  $\sigma$ " are good practice towards mastery of theorem proving.

Although this problem is in general undecidable, with the right strategy we can go a long way. To create a term of a given type, start with the placeholder \_ and recursively apply a combination of the following two steps:

- 1. If the type is of the form  $\sigma \to \tau$ , a possible inhabitant is an anonymous function, of the form fun  $x: \sigma \mapsto \_$ , where  $\_$  is a placeholder for a missing term of type  $\tau$ . Lean will mark  $\_$  as an error; if you hover over it in Visual Studio Code, a tooltip will show the missing term's type as well as any variables declared in the local context.
- 2. Given a type  $\sigma$  (which may be a function type), you can use any constant c or variable x of type  $\tau_1 \to \cdots \to \tau_n \to \sigma$  to build a term of type  $\sigma$ . For each argument, you need to put a placeholder, yielding c \_ . . . \_ or x \_ . . . \_ .

The placeholders can be eliminated recursively using the same strategy. As an example, we will apply the strategy to find a term of type

$$(\alpha \rightarrow \beta \rightarrow \gamma) \rightarrow ((\beta \rightarrow \alpha) \rightarrow \beta) \rightarrow \alpha \rightarrow \gamma$$

Initially, only step 1 is applicable, with

$$\sigma := \alpha \to \beta \to \gamma$$
 and  $\tau := ((\beta \to \alpha) \to \beta) \to \alpha \to \gamma$ 

(Recall that  $\rightarrow$  is right-associative.) This results in the term fun f  $\mapsto$  \_, which has the right type but has a placeholder left. Since the argument f has type  $\sigma$ , a function type, it makes sense to use the name f for it. Then we continue recursively with the placeholder, of type  $\tau$ . Again, only step 1 is possible, so we end up with the term fun f g  $\mapsto$  \_, where g has type ( $\beta \to \alpha$ )  $\to \beta$  and the placeholder has type  $\alpha \to \gamma$ . A third application of step 1 yields fun f g a  $\mapsto$  \_, where a has type  $\alpha$  and the placeholder has type  $\gamma$ .

At this point, step 1 no longer applies. Let us see if step 2 is possible. The context surrounding the placeholder contains the following variables:

$$f: \alpha \rightarrow \beta \rightarrow \gamma$$
,  $g: (\beta \rightarrow \alpha) \rightarrow \beta$ ,  $a: \alpha$ 

Recall that we are trying to build a term of type  $\gamma$ . The only variable we can use to achieve this is f: It takes two arguments and returns a value of type  $\gamma$ . So we replace the placeholder with the term f \_ \_, where the two new placeholders stand for the two missing arguments. Putting everything together, we now have the term f un f g a  $\mapsto$  f \_ \_.

Following f's type, the placeholders are of type  $\alpha$  and  $\beta$ , respectively. The first placeholder is easy to fill in, using step 2 again, by simply supplying a, of type  $\alpha$ , with no arguments. For the second placeholder, we apply step 2 with the variable g, which is the only source of  $\beta$ s. Since g takes an argument, we must supply a placeholder. This means our current term is fun f g a  $\mapsto$  f a (g \_).

We are almost done. The only placeholder left has type  $\beta \to \alpha$ , which is g's argument type. Applying step 1, we replace the placeholder with fun  $b \mapsto$  \_, where \_ has type  $\alpha$ . Here, we can simply supply a. Our final term is fun f g a  $\mapsto$  f a (g (fun  $b \mapsto$  a)).

The above derivation was tedious but deterministic: At each point, either step 1 or 2 was applicable, but not both. In general, this will not always be the case.

For some other types, we might encounter dead ends and need to backtrack. We might also fail altogether, with nowhere to backtrack to.

The key idea is that the term should be syntactically correct at all times. The only red underlining we should see in Visual Studio Code should appear under the placeholders. In general, a good principle for software development is to start with a program that compiles, perform the smallest change possible to obtain a new compiling program, and repeat until the program is complete.

## 1.5 Summary of Newly Introduced Lean Constructs

At the end of this and most other chapters, we include a brief summary of the constructs introduced in the chapter. Some syntaxes have multiple meanings, which will be introduced gradually. We refer to *The Lean 4 Manual* [1], the *Theorem Proving in Lean 4* [13] tutorial, and the mathlib documentation for details.

#### **Diagnostic Command**

#check checks and prints type of a term

#### **Declaration**

opaque declares an unspecified new constant or type

https://leanprover-community.github.io/mathlib4\_docs/

# **Chapter 2**

# **Programs and Theorems**

We continue studying the basics of Lean, focusing on programs and theorems, but without carrying out any proofs yet. In particular, we review how to define new types and functions and how to state their intended properties as theorems.

## 2.1 Type Definitions

A distinguishing feature of Lean's calculus of inductive constructions is its built-in support for inductive types. An *inductive type* is a type whose values are built by applying special constants called *constructors*. Inductive types are a concise way of representing acyclic data in a program. You may know them under some other, largely synonymous names, including algebraic data types, inductive data types, freely generated data types, recursive data types, and data types.

#### 2.1.1 Definition of Natural Numbers

The "Hello, World!" example of inductive types is the type Nat ( $\mathbb{N}$ ) of natural numbers. In Lean, it can be defined as follows:

```
inductive Nat : Type where
  | zero : Nat
  | succ : Nat → Nat
```

The first line announces to the world that we are introducing a new type called Nat, intended to represent the natural numbers. The second and third line declare two new constructors, Nat.zero: Nat and Nat.succ: Nat  $\rightarrow$  Nat, that can be used to build values of type Nat. Following an established convention in computer science and logic, counting starts at zero. The second constructor is what makes this inductive definition interesting—it requires an argument of type Nat to produce a value of type Nat. The terms

```
Nat.zero
Nat.succ Nat.zero
Nat.succ (Nat.succ Nat.zero)
:
```

denote the different values of type Nat—zero, its successor, its successor's successor, and so on. This notation is called *unary*—or *Peano*, after the logician Giuseppe

Peano. For an alternative explanation of Peano numbers in Lean (and some amazing video game graphics), see Kevin Buzzard's article "Can computers prove theorems?"

The general format of type declarations is

```
inductive type-name (params_1: type_1)...(params_k: type_k): Type where | constructor-name_1: constructor-type_1 \vdots | constructor-name_n: constructor-type_n
```

For the natural numbers, it is possible to tell Lean to use the familiar names 0, 1, 2, ..., and indeed the predefined  $\mathbb N$  type offers such syntactic sugar. But using the lengthier notations sheds more light on Lean's type definitions.

In the Lean file accompanying this chapter, the definition of Nat is located in a namespace, delimited by namespace MyNat and end MyNat, to restrict its scope to a portion of the file. After end MyNat, any occurrence of Nat, Nat.zero, or Nat.succ will refer to Lean's predefined type of natural numbers. Similarly, the entire file is put in the Love namespace to prevent name clashes with existing Lean libraries.

We can inspect an earlier definition at any point in Lean by using the #print command. For example, #print Nat within the MyNat namespace displays the following information:

```
inductive LoVe.MyNat.Nat : Type
number of parameters: 0
constructors:
LoVe.MyNat.Nat.zero : Nat
LoVe.MyNat.Nat.succ : Nat → Nat
```

The focus on natural numbers is one of the many features of this guide that reveal a bias towards computer science. Number theorists would be more interested in the integers  $\mathbb Z$  and the rational numbers  $\mathbb Q$ ; analysts would want to work with the real numbers  $\mathbb R$  and the complex numbers  $\mathbb C$ . But the natural numbers are ubiquitous in computer science and enjoy a very simple definition as an inductive type. They can also be used to build other types, as we will see in Chapter 12.

### 2.1.2 Definition of Arithmetic Expressions

Our next example uses the integers. If we were to specify a calculator program or a programming language, we would likely need to define a type to represent arithmetic expressions as an abstract syntax tree. The next example shows how this could be done in Lean:

```
inductive AExp : Type where
| num : \mathbb{Z} \to AExp
| var : String \to AExp
| add : AExp \to AExp \to AExp
| sub : AExp \to AExp \to AExp
| mul : AExp \to AExp \to AExp
| div : AExp \to AExp \to AExp
```

<sup>&</sup>lt;sup>1</sup>http://chalkdustmagazine.com/features/can-computers-prove-theorems/

Mathematically, this definition is equivalent to defining the type AExp inductively by the following formation rules:

- 1. For every integer i, the term AExp.num i is an AExp value. (Intuitively, the constructor AExp.num "boxes" an integer into an arithmetic expression.)
- 2. For every character string x, the term AExp. var x is an AExp value.
- 3. If e<sub>1</sub> and e<sub>2</sub> are AExp values, then so are AExp.add e<sub>1</sub> e<sub>2</sub>, AExp.sub e<sub>1</sub> e<sub>2</sub>, AExp.mul e<sub>1</sub> e<sub>2</sub>, and AExp.div e<sub>1</sub> e<sub>2</sub>.

The above definition is exhaustive. The only possible values for AExp are those built using formation rules 1 to 3. Moreover, AExp values built using different formation rules are distinct. These two properties of inductive types are captured by the motto "No junk, no confusion" due to Joseph Goguen.

It may be instructive to compare the concise Lean specification of AExp above with a Java program that achieves the same. The program consists of one interface and six classes that implement it, corresponding to the AExp type and its six constructors:

```
public interface AExp { }
public class Num implements AExp {
    public int num;
    public Num(int num) { this.num = num; }
}
public class Var implements AExp {
    public String var;
    public Var(String var) { this.var = var; }
}
public class Add implements AExp {
    public AExp left;
    public AExp right;
    public Add(AExp left, AExp right)
    { this.left = left; this.right = right; }
}
public class Sub implements AExp {
    public AExp left;
    public AExp right;
    public Sub(AExp left, AExp right)
    { this.left = left; this.right = right; }
}
public class Mul implements AExp {
    public AExp left;
```

```
public AExp right;
      public Mul(AExp left, AExp right)
      { this.left = left; this.right = right; }
 }
 public class Div implements AExp {
      public AExp left;
      public AExp right;
      public Div(AExp left, AExp right)
      { this.left = left; this.right = right; }
  }
In Python, the class declarations would be as follows:
 class AExp:
      pass
 class Num(AExp):
      def __init__(self, num):
          self.num = num
 class Var(AExp):
      def __init__(self, var):
          self.var = var
 class Add(AExp):
      def __init__(self, left, right):
          self.left = left
          self.right = right
 class Sub(AExp):
      def __init__(self, left, right):
          self.left = left
          self.right = right
 class Mul(AExp):
      def __init__(self, left, right):
          self.left = left
          self.right = right
 class Div(AExp):
      def __init__(self, left, right):
          self.left = left
          self.right = right
```

### 2.1.3 Definition of Lists

The next type we consider is that of finite lists:

```
inductive List (\alpha : Type) where 
 | nil : List \alpha 
 | cons : \alpha \rightarrow List \alpha \rightarrow List \alpha
```

The type is *polymorphic*: It is parameterized by a type  $\alpha$ , which we can instantiate with concrete types. For example, List  $\mathbb{Z}$  is the type of lists over integers, and List (List  $\mathbb{R}$ ) is the type of lists of lists of real numbers. The type constructor List takes a type as argument and returns a type. Like generics in Java and templates in C++, polymorphism is a mechanism that provides parameterized types.

The following commands display the constructors' types:

```
#check List.nil
#check List.cons
```

The output looks different from what we might expect:

```
List.nil : List ?m.2396 
 List.cons : ?m.2397 \rightarrow List ?m.2397 \rightarrow List ?m.2397
```

Even if we try to restrict ourselves to a fragment of Lean's language, Lean often exposes us to more advanced constructs in the output, such as ?m. 2396 and ?m. 2397 above. Our advice is to adopt a *sporty attitude*: Do not worry if you do not always understand everything the first time. Use your common sense and your imagination. And, above all, do not hesitate to ask.

Lean's built-in lists offer syntactic sugar for writing lists:

```
[] for List.nil x :: xs for List.cons x xs [x_1, ..., x_n] for x_1 :: ... :: x_n :: []
```

The :: operator, like all other binary operators, binds less tightly than function application. Thus, f x :: List.reverse ys is parsed as <math>(f x) :: (List.reverse ys). It is good practice to avoid needless parentheses. They can quickly impair readability. In addition, it is important to put spaces around infix operators, to suggest the right precedence.

Functional programmers often use names such as xs, ys, zs for lists, although l is also common in Lean. A list contains many elements, so a plural form is natural. A list of cats might be called cats; a list of lists of cats, catss. When a nonempty list is presented as a head and a tail, we usually write, say, x :: xs or cat :: cats.

### 2.2 Function Definitions

If all we want is to declare a function, we can use the opaque command (Section 1.2). But usually, we want to *define* the function's behavior, and for this we can use the def command. Because Lean's roots are in functional programming, functions are defined as mathematical expressions evaluating to a value, not as an imperative program modifying some state. Accordingly, recursion, rather than iteration with while or for, is the primary mechanism for traversing data.

#### 2.2.1 Recursion on Natural Numbers

Let us see on a simple example how recursion works. The Lean definition of Fibonacci numbers is as follows:

```
def fib : \mathbb{N} \to \mathbb{N}

| 0 => 0

| 1 => 1

| n + 2 => fib (n + 1) + fib n
```

The patterns on the left correspond to the argument of the function. Here, fib is declared with a single argument of type  $\mathbb{N}$ , so we pattern-match on a single natural number. A pattern is triggered when the input has the given form. For example, if the input is 1, then the pattern 1 matches it, and the corresponding right-hand side is evaluated. If the input is 5, then the pattern n + 2 is triggered, with n := 3. Then the right-hand side is evaluated with n set to 3, yielding fib 4 + fib 3.

The general format of recursive definitions is

The parameters  $params_1$  to  $params_m$  cannot be subjected to pattern matching, only the remaining arguments declared in type. If multiple patterns are provided on one line, they are separated by commas. Patterns may contain variables, which are then visible in the corresponding right-hand side, as well as constructors.

The basic arithmetic operations on natural numbers, such as addition, multiplication, and exponentiation, can be defined by recursion. Of course, they are already defined in Lean (as +, \*, and  $^{\circ}$ ), but they are useful exercises. We start with addition:

```
\begin{array}{lll} \text{def add} & : \; \mathbb{N} \; \to \; \mathbb{N} \; \to \; \mathbb{N} \\ & | \; \text{m, Nat.zero} & => \; \text{m} \\ & | \; \text{m, Nat.succ} \; \; \text{n} \; => \; \text{Nat.succ} \; \left( \; \text{add} \; \text{m} \; \text{n} \right) \end{array}
```

We pattern-match on two arguments at the same time, distinguishing the case where the second argument is zero and the case where it is nonzero. Each recursive call to add peels off one Nat.succ constructor from the second argument. Instead of Nat.zero and Nat.succ n, Lean also allows us to write 0 and n + 1 as syntactic sugar.

We can evaluate the result of applying add to numbers using #eval or #reduce:

```
#eval add 2 7
#reduce add 2 7
```

Both commands print 9 in Visual Studio Code. #eval employs an optimized interpreter, whereas #reduce uses Lean's inference kernel, which is less efficient.

It is good practice to provide a few tests each time we define a function, to ensure that it behaves as expected. You can even leave the #eval or #reduce calls in your Lean files as documentation.

The definition of multiplication is similar to that of addition:

```
def mul : \mathbb{N} \to \mathbb{N} \to \mathbb{N}
  | _, Nat.zero => Nat.zero
  | m, Nat.succ n => add m (mul m n)
```

The underscore (\_) stands for an unused variable. We could have put a name (e.g., m), but \_ documents our intentions better.

The #eval command below prints 14, as expected:

```
#eval mul 2 7
```

Exponentiation ("m to the power of n") can be defined in various ways. Our first proposal is structurally identical to the definition of multiplication:

```
def power : \mathbb{N} \to \mathbb{N} \to \mathbb{N}
  | _, Nat.zero => 1
  | m, Nat.succ n => mul m (power m n)
```

Since the first argument, m, remains unchanged in the recursive call, power m n, we can factor it out and put it next to the function's name, as a *parameter*, before the colon introducing the type of the function (excluding the parameter m):

```
def powerParam (m : \mathbb{N}) : \mathbb{N} \to \mathbb{N}
  | Nat.zero => 1
  | Nat.succ n => mul m (powerParam m n)
```

From the outside, there is no difference between the two definitions. In fact, we already saw this syntax for the type argument  $\alpha$  of the List constructor (Section 2.1).

Yet another definition is possible by first introducing a general-purpose iterator and then using it with the right arguments:

```
\begin{array}{lll} \text{def iter } (\alpha: \mathsf{Type}) \; (\mathsf{z}: \alpha) \; (\mathsf{f}: \alpha \to \alpha) : \mathbb{N} \to \alpha \\ \mid \; \mathsf{Nat.zero} \; \; = \!\!\!\! > \; \mathsf{z} \\ \mid \; \mathsf{Nat.succ} \; \mathsf{n} \; = \!\!\!\! > \; \mathsf{f} \; (\mathsf{iter} \; \alpha \; \mathsf{z} \; \mathsf{f} \; \mathsf{n}) \\ \\ \text{def powerIter } (\mathsf{m} \; \mathsf{n} \; : \; \mathbb{N}) \; : \; \mathbb{N} \; : = \\ & \mathsf{iter} \; \mathbb{N} \; \mathsf{1} \; (\mathsf{mul} \; \mathsf{m}) \; \mathsf{n} \end{array}
```

The iter function takes a type  $\alpha$ , a "zero" value z, a unary function f, and a natural number n, and computes

$$\underbrace{f\left(f\left(\cdots\left(f\right)z\right)\cdots\right)}_{\text{n times}}$$

It is an example of a higher-order function: a function that takes another function (here, f) as argument. In functional programming, functions are treated like any other objects and can be used as arguments or as the result of a function. Here, we use iter to compute m to the power of n as

Finally, notice that the definition of powerIter is not recursive. For nonrecursive functions without pattern matching, the syntax is simply

```
def name (params_1 : type_1) ... (params_m : type_m) : type := result
```

#### 2.2.2 Recursion on Arithmetic Expressions

Going back to the type of arithmetic expressions, if we wanted to implement an eval function in Java, we would probably add it as part of AExp's interface and implement it in each subclass. For Add, Sub, Mul, and Div, we would recursively call eval on the left and right objects.

In Lean, the syntax is very compact. We define a single function and use pattern matching to distinguish the six cases:

```
def eval (env : String \rightarrow \mathbb{Z}) : AExp \rightarrow \mathbb{Z}

| AExp.num i => i

| AExp.var x => env x

| AExp.add e_1 e_2 => eval env e_1 + eval env e_2

| AExp.sub e_1 e_2 => eval env e_1 - eval env e_2

| AExp.mul e_1 e_2 => eval env e_1 * eval env e_2

| AExp.div e_1 e_2 => eval env e_1 / eval env e_2
```

Notice that this function is higher-order: It takes a function, env, that represents an "environment" assigning values to the variables. The environment is used in the AExp.var case to evaluate a variable and is threaded through in the recursive cases (AExp.add, AExp.sub, AExp.mul, and AExp.div).

Perhaps you are worried about division by zero in the AExp.div case. Let us see what #eval has to say about it, using some environment that assigns 7 to all variables:

```
#eval eval (fun x \mapsto 7) (AExp.div (AExp.var "y") (AExp.num 0))
```

The output is 0. In Lean, division is conveniently defined as a total function that returns zero when the denominator is zero. For a lucid explanation of why this is not dangerous, see Buzzard's blog.<sup>2</sup>

#### 2.2.3 Recursion on Lists

Recursive functions on lists can be defined in a similar way:

The append function takes three arguments: a type  $\alpha$  and two lists of type List  $\alpha$ .

```
#eval append \mathbb{N} [3, 1] [4, 1, 5] #eval append _ [3, 1] [4, 1, 5]
```

By passing the placeholder \_, we leave it to Lean to infer the type  $\mathbb N$  from the type of the other two arguments.

To make the type argument  $\alpha$  implicit, we can put it in curly braces  $\{ \}$ :

```
def appendImplicit \{\alpha : \mathsf{Type}\} : \mathsf{List}\ \alpha \to \mathsf{List}\ \alpha \to \mathsf{List}\ \alpha | List.nil, ys => ys | List.cons x xs, ys => List.cons x (appendImplicit xs ys) #eval appendImplicit [3, 1]\ [4, 1, 5]
```

 $<sup>^2</sup> https://xenaproject.wordpress.com/2020/07/05/division-by-zero-in-type-theory-a-faq/$ 

The at sign (a) can be used to make the implicit arguments explicit. This is occasionally necessary to guide Lean's parser.

```
#eval @appendImplicit \mathbb{N} [3, 1] [4, 1, 5] #eval @appendImplicit \mathbb{N} [3, 1] [4, 1, 5]
```

We can use syntactic sugar in the definition, both in the patterns on the lefthand sides of => and in the right-hand sides:

```
def appendPretty \{\alpha : \mathsf{Type}\} : \mathsf{List}\ \alpha \to \mathsf{List}\
```

In Lean's standard library, the append function is an infix operator called ++. We can use it to define a function that reverses a list:

```
def reverse \{\alpha : \mathsf{Type}\} : \mathsf{List}\ \alpha \to \mathsf{List}\ \alpha

|\ [\ ]\ =>\ [\ ]

|\ x :: xs => reverse\ xs ++\ [x]
```

### 2.3 Theorem Statements

What makes Lean a proof assistant as well as a programming language is that we can state theorems about the types and constants we define and prove that they hold. In interactive theorem proving, the terms theorem, lemma, corollary, fact, property, and true statement are used more or less interchangeably. Similarly, propositions, logical formulas, and statements will all mean the same.

In Lean, propositions are simply terms of type Prop. This stands in contrast with first-order logic, where terms and formulas are traditionally distinguished in the syntax. A proposition that can be proved is called a theorem (or lemma, corollary, etc.); otherwise it is a nontheorem or false statement. Mathematicians sometimes use the term "proposition" as a synonym for theorem (e.g., "Proposition 3.14"), but in formal logic propositions can also be false.

Here are examples of true statements that can be made about the addition, multiplication, and list reversal operations defined in Section 2.2:

```
theorem add_comm (m n : N) :
   add m n = add n m :=
   sorry

theorem add_assoc (l m n : N) :
   add (add l m) n = add l (add m n) :=
   sorry

theorem mul_comm (m n : N) :
   mul m n = mul n m :=
   sorry

theorem mul_assoc (l m n : N) :
   mul (mul l m) n = mul l (mul m n) :=
   sorry
```

```
theorem mul_add (l m n : N) :
    mul l (add m n) = add (mul l m) (mul l n) :=
    sorry

theorem reverse_reverse {α : Type} (xs : List α) :
    reverse (reverse xs) = xs :=
    sorry

The general format is
    theorem name (params<sub>1</sub> : type<sub>1</sub>) ... (params<sub>m</sub> : type<sub>m</sub>) :
    statement :=
    proof
```

The := symbol separates the theorem's statement and its proof. The syntax of theorem is very similar to that of a def command without pattern matching, with statement instead of type and proof instead of result. In the examples above, we put the marker sorry as a placeholder for the actual proof. The marker is quite literally an apology to future readers and to Lean for the absence of a proof. It is also a reason to worry, until we manage to eliminate it. In Chapters 3 and 4, we will see how to achieve this.

The intuitive semantic of a theorem command with a sorry proof is, "This proposition should be provable, but I have not carried out the proof yet—sorry." Sometimes, we want to express a related idea, namely, "Let us assume this proposition holds." Lean provides the axiom command for this, which is often used in conjunction with opaque. For example:

```
opaque a : Z
opaque b : Z

axiom a_less_b :
  a < b</pre>
```

After the opaque commands, we have no information about a and b beyond their type. The axiom specifies a property they should have. The general format of the command is

```
axiom name (params_1 : type_1) ... (params_m : type_m) : statement
```

Axioms are dangerous, because they can lead to an inconsistent logic, in which we can derive False. For example, if we added a second axiom stating that a > b, we could easily derive False. The history of interactive theorem proving is paved with inconsistent axioms. One anecdote among many: At the 2020 edition of the Certified Programs and Proofs conference, a submitted paper was rejected due to a flawed axiom, from which one of the peer reviewers derived False.<sup>3</sup> Therefore, we will generally avoid axioms.

From Lean's point of view, a theorem with a sorry proof is effectively an axiom and can be used to play havoc with the logic's consistency. To prevent misunderstandings, it is better to use sorry only as a temporary measure, while developing a proof, and not as an alternative to the more explicit, and honest, axiom.

<sup>&</sup>lt;sup>3</sup>One of the authors of that submission reported that the axiom has now been revised and derived as a theorem. The entire formal development is now axiom-free. The revised paper has been accepted at the conference Computer Aided Verification 2020 [8].

# 2.4 Summary of Newly Introduced Lean Constructs

## **Diagnostic Commands**

#eval executes a term using an optimized interpreter

#print prints the definition of a constant

#reduce executes a term using Lean's inference kernel

#### **Declarations**

axiom states an axiom def defines a new constant

inductive introduces a type and its constructors collects declarations in a named scope

theorem states a theorem and its proof

#### **Proof Command**

stands for a missing proof or definition

# **Chapter 3**

# **Backward Proofs**

A tactic is a procedure that operates on the goal—the proposition to prove—and either fully proves it or produces new subgoals, or fails. When we state a theorem, the theorem statement is the initial goal. A proof is complete once all (sub)goals have been eliminated using suitable tactics. Most of the tactics described here are documented in more detail in Chapter 5 of Theorem Proving in Lean 4 [13].

Tactics are a *backward* proof mechanism. They start from the goal and work backwards towards the already proved theorems. Consider the theorems a,  $a \to b$ , and  $b \to c$  as well as the goal  $\vdash c$ . (Goals, like typing judgments, are identified by the symbol  $\vdash$ .) An informal backward proof is as follows:

```
To prove c, by b \to c it suffices to prove b. 
To prove b, by a \to b it suffices to prove a. 
To prove a, we use a.
```

The telltale sign of a backward proof is the phrase "it suffices to." Notice how we progress from one *goal* to another ( $\vdash$  c,  $\vdash$  b,  $\vdash$  a) until no goal is left to prove. By contrast, a *forward* proof would start from the theorem a and progress, one *theorem* at a time, towards the desired theorem c:

```
From a and a \rightarrow b, we have b. From b and b \rightarrow c, we have c, as desired. \Box
```

A forward proof only manipulates theorems, not goals. We will study forward proofs more deeply in Chapter 4.

Informal proofs are sometimes written in a mixture of both styles. This is manageable as long as the backward steps are clearly identified, by such wording as "to prove ..., it suffices to prove ...." It may help to put the  $\vdash$  sign in front of goals as a reminder that these claims have not been proved yet.

Another format for expressing proofs you may be familiar with is natural deduction. A natural deduction derivation corresponding to the above proof is given below:

When read from top to bottom, the derivation corresponds to a forward proof; when read from bottom to top, it corresponds to a backward proof.

### 3.1 Tactic Mode

In Chapter 2, whenever a proof was required, we simply used a sorry placeholder. For a tactical proof, we will now write by to enter *tactic mode*. In this mode, we can invoke a sequence of tactics.

Tactics operate on the goal, which consists of the proposition Q that we want to prove and of a local context C. The local context consists of variable declarations of the form  $x:\sigma$  and hypotheses of the form h:P. We write  $C\vdash Q$  to denote a goal, where C is a list of variables and hypotheses and Q is the goal's target.

To make things more concrete, consider the following Lean example:

```
theorem fst_of_two_props :
    ∀a b : Prop, a → b → a :=
    by
        intro a b
        intro ha hb
        apply ha
```

Note that the implication arrow  $\rightarrow$  is right-associative; this means that  $a \rightarrow b \rightarrow a$  is the same as  $a \rightarrow (b \rightarrow a)$ . Intuitively speaking, the statement has the meaning "a implies that b implies a," or equivalently "a and b imply a."

We will review this proof line by line. The

```
by
```

keyword indicates that we enter tactic mode, in which we can specify a proof by means of tactics. The tactics then follow, one per line and with the same indentation for all.

If we put the cursor on the by keyword, Visual Studio Code reports that the current goal is simply the original theorem statement:

```
\label{eq:lambda} \vdash \ \forall a \ b : \ \mathsf{Prop}, \ a \to b \to a Next, the tactic \quad \mathsf{intro} \ a \ b
```

tells Lean to fix two free variables, a and b, corresponding to the two bound variables of the same names. The tactic mimics how mathematicians work on paper: To prove a  $\forall$ -quantified proposition, it suffices to prove it for some arbitrary but fixed values of the bound variables. The goal becomes

```
ab: Prop \vdash a \rightarrow b \rightarrow a
```

where the propositions a and b are declared in the goal's local context, on the left of the  $\vdash$  symbol. We often use the same names for the free variables as for bound variables, as we did here, but this is not mandatory. In general, it is good practice to use unique names and avoid shadowing existing variables.

Instead of a single call to intro with two variable names, we could also have called intro twice, as intro a followed by intro b. In general, each intro invocation with n arguments extracts the next n quantified variables.

The intro tactic is not limited to quantified variables. In our proof script,

3.1. Tactic Mode 25

```
intro ha hb
```

tells Lean to move the assumptions a and b (from a  $\rightarrow$  b  $\rightarrow$ ) to the local context and to call these hypotheses ha and hb.

Indeed, to prove an implication, it suffices to take its left-hand side as hypothesis and prove its right-hand side. Thus the goal becomes

```
ab: Prop, ha: a, hb: b \vdash a
```

It is customary to prefix hypothesis names with h.

Finally, the tactic

```
apply ha
```

tells Lean to match the hypothesis a, called ha, against the goal  $\vdash$  a. Since a is syntactically equal to a, we have a match. This completes the proof.

Informally, in a style reminiscent of pen-and-paper mathematics, we could write the proof as follows:

Let a and b be propositions.

Assume (ha) a and (hb) b are true.

To prove a, we use hypothesis ha.

(Mathematicians would probably use numeric tags such as (1) and (2) for the hypotheses instead of informative names.)

Going back to the Lean proof, we can avoid the intro invocations by declaring the variables and hypotheses as parameters of the theorem, as follows:

```
theorem fst_of_two_props_params (a b : Prop) (ha : a) (hb : b) :
  a :=
  by apply ha
```

The goal is

```
ab: Prop, ha: a, hb: b \vdash a
```

All the parameters of the theorem are immediately available in the context. The goal is proved by apply ha, like in our previous example.

Here is an example with multiple applys in sequence:

```
theorem prop_comp (a b c : Prop) (hab : a \rightarrow b) (hbc : b \rightarrow c) : a \rightarrow c := by  
   intro ha  
   apply hbc  
   apply hab  
   apply ha
```

Putting on our mathematician's hat, we can verbalize the last proof as follows:

Assume (ha) a is true.

To prove c, by hypothesis hbc it suffices to prove b.

To prove b, by hypothesis hab it suffices to prove a.

To prove a, we use hypothesis ha.

### 3.2 Basic Tactics

The intro and apply tactics are the staples of tactical proofs. Other basic tactics include exact, assumption, rfl, and ac\_rfl. These tactics can go a long way, if we are patient enough to carry out the reasoning using them, without appealing to stronger proof automation. They can also be used to solve various logic puzzles.

Below, the thin, large square brackets [] enclose optional syntax.

#### intro

```
intro [name_1 \dots name_n]
```

The intro tactic moves the leading  $\forall$ -quantified variable or the leading assumption  $a \rightarrow$  from the goal's target to the local context. The tactic takes as an optional argument the name to give to the variable or to the assumption in the context, overriding the default name. If multiple names are supplied (i.e., n > 1), multiple variables or assumptions are moved.

intro always preserves provability and is therefore safe.

#### rfl

The rfl tactic proves goals of the form  $\vdash$  l = r, where the two sides l and r are syntactically equal *up to computation*. Computation means first and foremost the expansion of definitions but also the reduction of an application of an anonymous function to an argument, and more. These *conversions* have traditional names. The main conversions are listed below together with examples, in a global context containing def double (n :  $\mathbb{N}$ ) :  $\mathbb{N}$  := n + n:

```
α-conversion (\operatorname{fun} x \mapsto f x) = (\operatorname{fun} y \mapsto f y)

β-conversion (\operatorname{fun} x \mapsto f x) = f a

δ-conversion double f = f a

ζ-conversion (\operatorname{let} n : \mathbb{N} := 2; n + n) = f a

η-conversion (\operatorname{fun} x \mapsto f x) = f a

ι-conversion (\operatorname{fun} x \mapsto f x) = f a
```

Applying conversions repeatedly as left-to-right rewrite rules is called *reduction*; applying a conversion once in reverse is called *expansion*.

Briefly, to prove  $\vdash$  l = r, the rfl tactic expands definitions in l and r and performs  $\beta$ -reduction and other reductions. It succeeds if l and r become syntactically identical at some point during this reduction process. Often, rfl succeeds where a mathematician would say "by definition."

#### apply

```
apply theorem-or-hypothesis
```

The apply tactic matches the goal's target with the conclusion of the specified theorem or hypothesis and adds the theorem or hypothesis's assumptions as new goals. The matching is performed up to computation.

We must invoke apply with care, because it can transform a provable goal into an unprovable subgoal. For example, if the goal is  $\vdash$  True and we apply the theorem False  $\rightarrow$  True, the conclusion True is matched against the goal's target True, and we end up with the unprovable subgoal  $\vdash$  False. We say that apply is unsafe.

#### exact

```
exact theorem-or-hypothesis
```

The exact tactic matches the goal's target with the specified theorem or hypothesis, proving the goal. We can often use apply in such situations, but exact communicates our intentions better.

#### assumption

The assumption tactic finds a hypothesis from the local context that matches the goal's target and applys it to prove the goal.

#### ac\_rfl

The ac\_rfl is similar to rfl, but it can be used to reason up to associativity (e.g., (a + b) + c = a + (b + c)) and commutativity (e.g., a + b = b + a). This works for binary operations that are registered as associative and commutative, such as + and \* on arithmetic types, and  $\cup$  and  $\cap$  on sets. We will see an example in Section 3.6.

#### sorry

The sorry proof command we encountered in Chapter 2 can be used at any point in a tactical proof, as a tactic. It "proves" the current goal without actually proving it. Use with care.

# 3.3 Reasoning about Connectives and Quantifiers

Before we learn to reason about natural numbers, lists, or other data types, we must first learn to reason about Lean's logical connectives and quantifiers. Let us start with a simple example: commutativity of conjunction  $(\land)$ .

```
theorem And_swap (a b : Prop) :
  a ∧ b → b ∧ a :=
  by
   intro hab
  apply And.intro
  apply And.right
  exact hab
  apply And.left
  exact hab
```

At this point, we recommend that you move the cursor over the example in Visual Studio Code, to see the sequence of proof states. By putting the cursor on or immediately after each command, you can see the effect of the command on that line. For the last line, Lean simply reports "Goals accomplished," meaning that no subgoals remain to be proved.

The proof is a typical intro-apply-exact mixture. It uses the theorems

```
And.intro : ?a \rightarrow ?b \rightarrow ?a \land ?b
And.left : ?a \land ?b \rightarrow ?a
And.right : ?a \land ?b \rightarrow ?b
```

where the question marks (?) indicate variables that can be instantiated—for example, by matching the goal's target against the conclusion of a theorem. These are called *metavariables*.

The three theorems above are the introduction rule and the two elimination rules associated with conjunction. An *introduction rule* for a logical symbol (e.g.,  $\land$ ) is a theorem whose conclusion has that symbol as the outermost symbol. Dually, an *elimination rule* has the symbol in an assumption. For each logical symbol, the introduction rules tell us *how to prove* a proposition with that symbol as the outermost position. By contrast, the elimination rules tell us *how we must have proved* such a proposition.

In the above proof, we apply the introduction rule for  $\land$  to prove the goal  $\vdash$  b  $\land$  a, and we apply the two elimination rules to extract b and a from the hypothesis a  $\land$  b. It may sound strange that we "eliminate"  $\land$  in  $\vdash$  b  $\land$  a by using a so-called introduction rule. The terminology is backward because our proofs are backward.

Question marks can also arise in goals. They indicate variables that can be instantiated arbitrarily. In the middle of the proof above, right after the tactic apply And.right, we have the goal

```
ab: Prop, hab: a \land b \vdash ?left.a \land b
```

where ?left.a is a metavariable. The tactic exact hab matches ?left.a (in the target) with a (in hab). The procedure used to instantiate variables to make two terms syntactically equal is called *unification*. *Matching* is a special case of unification where one of the two terms contains no variables, like here.

Incidentally, whenever metavariables appear in goals, additional subgoals will also appear, with the type of each metavariable as a subgoal (e.g.,  $\vdash$  Prop). These confusing subgoals are merely a reminder that we will need to instantiate the metavariables with terms of the right type. We can usually ignore these subgoals. As soon as the metavariable is instantiated (typically when we solve another subgoal), its associated subgoal will also go away.

In Lean, unification is performed up to computation. For example, the terms (fun  $x \mapsto ?m$ ) a and b can be unified by taking ?m := b, because (fun  $x \mapsto b$ ) a and b are syntactically equal up to  $\beta$ -conversion.

The following is an alternative proof of the theorem And\_swap:

```
theorem And_swap_braces :
    ∀a b : Prop, a ∧ b → b ∧ a :=
    by
    intro a b hab
    apply And.intro
    { exact And.right hab }
    { exact And.left hab }
```

The theorem is stated differently, with a and b as  $\forall$ -quantified variables instead of parameters of the theorem. Logically, this is equivalent, but in the proof we must then introduce a and b in addition to hab.

Another difference is the use of curly braces  $\{ \}$ . When we face two or more goals to prove, it is generally good style to put each proof in its own block enclosed in curly braces. The  $\{ \}$  tactic combinator focuses on the first subgoal; the tactic inside must prove it. In our example, the apply And.intro tactic creates two subgoals,  $\vdash$  b and  $\vdash$  a.

A third difference is that we now apply, by juxtaposition, And.right and And. left directly to the hypothesis  $a \land b$  to obtain b and a, respectively, instead of waiting for the theorems' assumptions to emerge as new subgoals. This is a small forward step in an otherwise backward proof. The same syntax is used both to discharge (i.e., prove) a hypothesis and to instantiate  $a \lor d$  quantifier. One benefit of this approach is that we avoid the potentially confusing ?left.a metavariable.

In the next example, we use juxtaposition to instantiate a  $\forall$  quantifier:

```
theorem f5_if (h : ∀n : N, f n = n) :
f 5 = 5 :=
by exact h 5
```

If h is the theorem  $\forall$ n, f n = n, then h 5 is the theorem f 5 = 5.

The introduction and elimination rules for disjunction ( $\lor$ ) are as follows:

```
Or.inl: \forallb: Prop, ?a \rightarrow ?a \vee b
Or.inr: \forallb: Prop, ?a \rightarrow b \vee ?a
Or.elim: ?a \vee ?b \rightarrow (?a \rightarrow ?c) \rightarrow (?b \rightarrow ?c) \rightarrow ?c
```

The  $\forall$  quantifiers in Or.inl ("introduction left") and Or.inr ("introduction right") can be instantiated directly by applying the theorem name to the value we want to instantiate with, via simple juxtaposition. Thus, Or.inl False corresponds to the theorem ?a  $\rightarrow$  ?a  $\vee$  False. This is the forward style.

Alternatively, we can invoke apply Or.inl on a goal of the form  $\dots \vdash c \lor d$ . This sets ?a := c and ?b := d in the theorem. The new subgoal is  $\dots \vdash c$ . This is the backward style.

Both Or.inl and Or.inr are unsafe: If you apply the wrong one of the two, or either of them too early in a proof, you might end up with an unprovable subgoal. This is easy to see if you consider the provable goal  $\vdash$  True  $\lor$  False: applying Or.inr yields the unprovable subgoal  $\vdash$  False.

The Or.elim rule may seem counterintuitive at first glance. In essence, it states that if we have a  $\vee$  b, then to prove an arbitrary c, it suffices to prove c when a holds and when b holds. You can think of  $(?a \rightarrow ?c) \rightarrow (?b \rightarrow ?c) \rightarrow ?c$  as a clever trick to express the meaning of disjunction using only implication.

The introduction and elimination rules for equivalence  $(\leftrightarrow)$  are as follows:

```
Iff.intro: (?a \rightarrow ?b) \rightarrow (?b \rightarrow ?a) \rightarrow (?a \leftrightarrow ?b)
Iff.mp: (?a \leftrightarrow ?b) \rightarrow ?a \rightarrow ?b
Iff.mpr: (?a \leftrightarrow ?b) \rightarrow ?b \rightarrow ?a
```

The introduction and elimination rules for existential quantification  $(\exists)$  are

```
Exists.intro: \forall w, (?Pw \rightarrow (\exists x, ?Px))
Exists.elim: (\exists x, ?Px) \rightarrow (\forall a, ?Pa \rightarrow ?c) \rightarrow ?c
```

The introduction rule for  $\exists$  can be used to instantiate an existential quantifier with a *witness*—a value for which the quantifier's body is true. For example:

```
theorem Exists_double_iden :
    ∃n : N, double n = n :=
    by
    apply Exists.intro o
    rfl
```

Again, we instantiate a  $\forall$  quantifier in a forward fashion: Exists.intro o is the theorem ?P o  $\rightarrow$  ( $\exists x$ , ?P x). The rule is unsafe: Choosing the wrong witness for x will result in an unprovable goal. For example, if the goal is  $\vdash \exists n$ , n > 5 and we take 3 as the witness, we end up with the unprovable subgoal  $\vdash 3 > 5$ .

The elimination rule for  $\exists$  is reminiscent of that for  $\lor$ . Indeed, a fruitful way to think of a quantification  $\exists n$ , ?P n is as a possibly infinitary disjunction  $?P o \lor ?P 1 \lor \cdots$ . Similarly,  $\forall n$ , ?P n can be thought of as  $?P o \land ?P 1 \land \cdots$ .

For truth (True), there is only an introduction rule:

```
True.intro: True
```

Truth holds no information whatsoever. If it appears as a hypothesis, it is completely useless, and there is no elimination rule that can extract any information from it. The clear tactic, described in Section 3.8 below, can be used to remove such useless hypotheses.

Dually, for falsehood (False), there is only an elimination rule:

```
False.elim : False \rightarrow ?a
```

There is no way to prove falsehood, but if we somehow have it from somewhere (e.g., from a hypothesis), then we can derive ?anything.

In fact, negation (Not) is defined in terms of implication and falsehood:  $\neg$  a abbreviates  $a \rightarrow False$ . Intuitively, these mean the same. We can think of "not a" as saying that "a would imply something absurd (hence not a)."

Lean's logic is classical, with support for the law of excluded middle and proof by contradiction:

```
Classical.em : \forall a : Prop, a \lor \neg a Classical.byContradiction : (\neg ?a \to False) \to ?a
```

Implication  $(\rightarrow)$  and universal quantification  $(\forall)$  are the proverbial dogs that did not bark. They do not have introduction or elimination rules. Instead, for both of them, the intro tactic is the introduction principle, and application (as in the  $\lambda$ -calculus) is the elimination principle. For example, given the theorems hab:  $a \rightarrow b$  and ba:  $a \rightarrow b$ :  $a \rightarrow$ 

For proving logic puzzles involving connectives and quantifiers, we advocate a "mindless," "video game" style of reasoning that relies mostly on basic tactics such as intro and apply. Here are some strategies that often work:

- If the goal's target is an implication  $P \to Q$ , invoke intro hP to move P into your hypotheses: ..., hP :  $P \vdash Q$ .
- If the goal's target is a universal quantification  $\forall x : \sigma$ , Q, invoke intro x to move x into the local context: ...,  $x : \sigma \vdash Q$ .

- Look for a theorem or hypothesis whose conclusion has the same shape as the goal's target (possibly containing variables that can be matched), and apply it. For example, if the goal's target is Q and you have a theorem or hypothesis of the form hPQ:  $P \rightarrow Q$ , try apply hPQ.
- A negated goal ⊢ ¬P is syntactically equal to ⊢ P → False up to computation, so you can invoke intro hP to produce the subgoal hP: P ⊢ False. Expanding negation's definition by invoking rw [Not] (described in Section 3.5) is often a good strategy.
- Sometimes you can make progress by replacing the goal by False, by entering apply False.elim. As the next step, you would typically apply a theorem or hypothesis of the form  $P \to False$  or  $\neg P$ .
- When you face several choices (e.g., between Or.inl and Or.inr), remember which choices you have made, and backtrack when you reach a dead end or feel that you are not making any progress.
- If you suspect that you might have reached a dead end, check whether the goal actually is provable under the given assumptions. Even if you started with a provable theorem statement, the current goal might be unprovable (e.g., if you applied unsafe rules).

# 3.4 Reasoning about Equality

Equality (=) is also a basic logical constant. It is characterized by the following introduction and elimination rules:

```
Eq.refl : \forall a, a = a

Eq.symm : ?a = ?b \rightarrow ?b = ?a

Eq.trans : ?a = ?b \rightarrow ?b = ?c \rightarrow ?a = ?c

Eq.subst : ?a = ?b \rightarrow ?P ?a \rightarrow ?P ?b
```

The first three theorems are introduction rules specifying that = is an equivalence relation. The fourth theorem is an elimination rule that allows us to replace equals for equals in an arbitrary context, represented by the metavariable ?P.

An example will show some of these rules in action. Below, we apply Eq.trans and Eq.symm to prove a = c using the equations a = b and c = b:

```
theorem Eq_trans_symm {α : Type} (a b c : α)
    (hab : a = b) (hcb : c = b) :
    a = c :=
    by
    apply Eq.trans
    { exact hab }
    { apply Eq.symm
        exact hcb }
```

Since rewriting in this way is such a common operation, Lean provides a rw tactic to achieve the same result. The tactic will also notice if rfl is applicable:

```
theorem Eq_trans_symm_rw \{\alpha : Type\} (a b c : \alpha)
(hab : a = b) (hcb : c = b) :
a = c :=
```

```
by
  rw [hab]
  rw [hcb]
```

A note on parsing: Equality binds more tightly than the logical connectives. Thus,  $a = b \wedge c = d$  is read as  $(a = b) \wedge (c = d)$ .

# 3.5 Rewriting Tactics

The rewriting tactic rw and its relative simp replace equals for equals. They use equations as left-to-right rewrite rules, replacing occurrences of the left-hand side by the right-hand side.

By default, they operate on the goal's target, but they can also be used to rewrite hypotheses specified using the at keyword:

```
at h_1 	ext{...} h_n rewrites the specified hypotheses at * rewrites all hypotheses and the target
```

rw

```
rw [theorem-or-constant<sub>1</sub>, ..., theorem-or-constant<sub>n</sub>] [at position]
```

The rw tactic rewrites the goal using one or more equations as left-to-right rewrite rules. It searches for the first subterm that matches any of the equations' left-hand side; once found, all occurrences of that subterm are replaced by the right-hand side of the equation. If the equation contains variables, these are instantiated as necessary. To use a theorem in reverse, as a right-to-left rewrite rule, put a short left arrow ( $\leftarrow$ ) in front of the theorem's name.

Thus, given the theorem hg:  $\forall x$ , g x = f x and the goal  $\vdash h (f a) (g b) (g c)$ , the tactic rw [hg] produces the subgoal  $\vdash h (f a) (f b) (g c)$ , whereas rw [ $\leftarrow$ hg] produces the subgoal  $\vdash h (g a) (g b) (g c)$ .

Instead of a theorem, we can also specify the name of a constant. This will attempt to use one of the constant's defining equations as rewrite rules.

#### simp

```
simp [at position]
```

The simp tactic rewrites the goal using a standard set of rewrite rules, called the simp set, exhaustively. Each equation in the simp set is used as a left-to-right rewrite rule. The simp set contains various rules about predefined symbols (e.g., arithmetic and list operators) and can be extended by putting the <code>@[simp]</code> attribute on suitable theorems.

```
simp [theorem-or-constant_1, ..., theorem-or-constant_n] [at position]
```

For the above simp variant, the specified theorems are temporarily added to the simp set. In the theorem list, an asterisk (\*) can be use to represent all hypotheses. The minus sign (-) in front of a theorem name temporarily removes the theorem from the simp set. A powerful incantation that both simplifies the hypotheses and uses the result to simplify the goal's target is simp [\*] at \*.

Given the theorem  $hg : \forall x, g x = f x$  and the goal  $\vdash h (f a) (g b) (g c)$ , the tactic simp [hg] produces the subgoal  $\vdash h (f a) (f b) (f c)$ , where both g b and g c have been rewritten. Instead of a theorem, we can also specify the name of a constant. This temporarily adds the constant's defining equations to the simp set.

At this point, you might wonder, "So what does simp do exactly?" Of course, you could study the source code or look up the scientific literature. But this might not be the most efficient use of your time. In truth, even expert users of proof assistants do not fully understand the behavior of the tactics they use daily. The most successful users adopt a relaxed, sporty attitude, trying tactics in sequence and studying the emerging subgoals, if any, to see if they are on the right path.

As you keep on using simp and other tactics, you will develop some intuition about what kinds of goals they work well on. This is one of the many reasons why interactive theorem proving can be learned only by doing. Often, you will not understand exactly what Lean does—why a tactic succeeds, or fails. Theorem proving can be very frustrating at times. The advice printed in large, friendly letters on the cover of *The Hitchhiker's Guide to the Galaxy* applies here: DON'T PANIC.

# 3.6 Proofs by Mathematical Induction

The induction tactic performs structural induction on a value of an inductive type. Structural induction means that the induction follows the structure of the inductive type. For natural numbers constructed from Nat.zero and Nat.succ, structural induction corresponds to standard mathematical induction: To prove p n, it suffices to prove p o and  $\forall k$ , p k  $\rightarrow$  p (k + 1). Equipped with induction, we can reason about the addition and multiplication operations we defined by recursion in Section 2.2.

Addition is defined by recursion on its second argument. We will prove two theorems, add\_zero and add\_succ, that give us alternative equations that recurse on the first argument. We start with add\_zero:

```
theorem add_zero (n : N):
  add o n = n :=
by
  induction n with
  | zero => rfl
  | succ n' ih => simp [add, ih]
```

The induction tactic is followed by two subproofs identified by the constructor they correspond to. In addition, any variables or hypotheses specific to the subproof can be named explicitly after the name of the constructor.

The first case, labeled zero, corresponds to the base case  $\vdash$  add 0 0 = 0. The second case, labeled succ, corresponds to the induction step

```
n': \mathbb{N}, ih: add o n' = n' \vdash add o (Nat.succ n') = Nat.succ n'
```

The names n' and ih specified after succ are the names we give to the argument of Nat.succ and the induction hypothesis, respectively. Other names are possible, but it is conventional to call the induction hypothesis ih.

We can keep on proving theorems by structural induction:

```
theorem add_succ (m n : \mathbb{N}):
 add (Nat.succ m) n = Nat.succ (add m n) :=
    induction n with
    lzero
                 => rfl
    | succ n' ih => simp [add, ih]
theorem add comm (m n : \mathbb{N}):
 add m n = add n m :=
 by
    induction n with
    | zero => simp [add, add zero]
    | succ n' ih => simp [add, add_succ, ih]
theorem add assoc (l m n : \mathbb{N}):
 add (add l m) n = add l (add <math>m n) :=
 by
    induction n with
    | zero => rfl
    | succ n' ih => simp [add, ih]
```

Once we have proved that a binary operator is commutative and associative, it is a good idea to let Lean's automation, notably ac\_rfl, know about this. The following commands achieve this for add:

```
instance IsAssociative_add : IsAssociative N add :=
  { assoc := add_assoc }

instance IsCommutative_add : IsCommutative N add :=
  { comm := add_comm }
```

(The instance mechanism will be explained in Chapter 5.) The following example uses the ac\_rfl tactic to reason up to associativity and commutativity of add:

Here are a few hints on how to carry out proofs by induction:

- It is usually beneficial to perform induction following the structure of the definition of one of the functions appearing in the goal. In particular, if a function is defined by recursion on its *n*th argument, it can make sense to perform the induction on that argument.
- If the base case of an induction is difficult, this is often a sign that the wrong variable was chosen or that some theorems should be proved first.

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## 3.7 Induction Tactic

#### induction

The induction tactic performs structural induction on the specified term. This gives rise to as many subgoals as there are constructors in the definition of the term's type. Induction hypotheses are available as hypotheses in the subgoals corresponding to recursive constructors (e.g., Nat.succ or List.cons). The optional names names<sub>1</sub>, ..., names<sub>n</sub> are used for any emerging variables or hypotheses.

# 3.8 Cleanup Tactics

The following tactics help us clean up the goal. So far they have not been needed, but they can be helpful during proof exploration.

#### clear

```
clear variable-or-hypothesis<sub>1</sub>... variable-or-hypothesis<sub>n</sub>
```

The clear tactic removes the specified variables and hypotheses, as long as they are not used anywhere else in the goal.

#### rename

```
rename type-of-variable-or-proposition-of-hypothesis => new-name
```

The rename tactic changes the name of a variable or hypothesis.

# 3.9 Summary of Newly Introduced Lean Constructs

#### **Attribute**

```
adds a theorem to the simp set
```

#### **Proof Command**

by initiates a tactical proof

#### **Tactics**

proves l = r up to associativity and commutativity ac\_rfl matches the goal's target against the theorem's conclusion apply proves the goal using a hypothesis assumption removes a variable or hypothesis from the goal clear proves the goal using the specified theorem exact performs structural induction on a variable of an inductive type induction moves  $\forall$ -quantified variables into the goal's hypotheses intro renames a variable or hypothesis rename proves l = r up to computation rfl rewrites once using the given theorem as a rewrite rule rewrites using a set of preregistered rewrite rules exhaustively simp sorry stands for a missing proof

#### **Tactic Combinator**

 $\{ \ \dots \ \}$  focuses on the first subgoal; needs to prove that subgoal

# **Chapter 4**

# **Forward Proofs**

Tactical proofs work backwards. They start from the goal and reduce it to existing theorems and hypotheses. Often it makes sense to work in a forward fashion: to start with existing theorems and hypotheses and proceed step by step towards our goal. Structured proofs are a style that supports this kind of reasoning. Tactical proofs tend to be easier to write but harder to read. Most users combine the two styles, using whichever seems the most appropriate for the situation. The higher readability of structured proofs make them popular with some users.

Structured proofs are syntactic sugar sprinkled over Lean's proof terms. They are built using keywords such as assume, have, let, and show that mimic penand-paper proofs. All Lean proofs, whether tactical or structured, are reduced internally to proof terms. We have seen some specimens already, in Chapter 3: Given hypotheses ha: a and hab:  $a \rightarrow b$ , the term hab ha is a proof term for the proposition b, and we write hab ha: b. The names of theorems and hypotheses, such as ha and hab, are also proof terms. Pushing this idea further, given hbc:  $b \rightarrow c$ , we can build the proof term hbc (hab ha) for the proposition c. We can think of hab as a function that converts a proof of a to a proof of b, and similarly for hbc.

Structured proofs are the default in Lean. They can be used outside tactic mode. To enter tactic mode, we must use the by command.

The concepts covered here are described in more detail in Chapters 2 to 4 of *Theorem Proving in Lean* 4 [13]. Nederpelt and Geuvers's textbook [22] and Van Raamsdonk's lecture notes [26] are other useful references.

#### 4.1 Structured Proofs

As a first example, consider the following structured proof:

```
theorem fst_of_two_props:
\forall a \ b : Prop, \ a \rightarrow b \rightarrow a :=
fix \ a \ b : Prop
assume \ ha : a
assume \ hb : b
show \ a \ from
ha
```

Each variable bound by a  $\forall$  quantifier and each assumption of an implication is introduced explicitly in the proof using the fix and assume commands. Several variables can be introduced simultaneously. We will often omit the types of variables,

especially when they can be guessed from their names; however, we will always spell out the propositions and put them one per line, to increase readability—which is, after all, one of the main potential advantages of the structured style. The show ... from command at the end repeats the proposition to prove, for the sake of readability, and gives the proof after the keyword from. The goal at this point is a b: Prop, ha: a, hb:  $b \vdash a$ .

Informally, we could write the proof as follows:

```
Fix some propositions a and b.

Assume (ha) a and (hb) b are true.

We must show a. This follows trivially from ha.
```

Some authors would insert qualifiers such as "arbitrary but fixed" in front of "propositions," or they would write "Let a and b be some propositions." All these variants are equivalent. And instead of " $\Box$ ," we could write "QED" to conclude the proof.

The Lean proof above is atypical in that the goal's target appears among the hypotheses. Usually, we must perform some intermediate reasoning steps, essentially of the form "from so-and-so, we have such-and-such." In Lean, each intermediate step takes the form of a have command, as in the following example:

```
theorem prop_comp (a b c : Prop) (hab : a \rightarrow b) (hbc : b \rightarrow c) : a \rightarrow c := assume ha : a have hb : b := hab ha have hc : c := hbc hb show c from hc
```

Informally:

```
Assume (ha) a is true.

From ha and hab, we have (hb) b.

From hb and hbc, we have (hc) c.

We must show c. This follows trivially from hc.
```

Notice that this is a forward proof: It progresses one theorem at a time from the hypothesis a to the desired theorem c. Modus ponens ("from a and a  $\rightarrow$  b, derive b") is expressed by simple juxtaposition (e.g., hab ha).

In general, the fix-assume-show skeleton repeats the theorem's statement. We often name the fixed variables after the bound variables from the goal, as we did here, but this is not mandatory. Moreover, it is good practice to avoid shadowing existing variables. Between the last fix or assume and the show, we can have as many have commands as we want, depending on how detailed we want the argument to be. Details can increase readability, but providing too many details can overwhelm the reader.

The have command has a similar syntax to theorem but appears inside a structured proof. We can also think of a have as a definition. In have hb: b:= hab ha,

the right-hand side hab ha is a proof term for b, and the left-hand side hb is defined as a synonym for that proof term. From then on, hb and hab ha can be used interchangeably. Since hb and hc are used only once and their proofs are very short, experts would tend to inline them, replacing hc by hbc hb and then hb by hab ha, yielding

```
theorem prop_comp_inline (a b c : Prop) (hab : a \rightarrow b) (hbc : b \rightarrow c) : a \rightarrow c := assume ha : a show c from hbc (hab ha)
```

A typical structured proof has the following fix-assume-have-show format:

```
theorem hr: \forall (c_1:\sigma_1) \ldots (c_l:\sigma_l), P_1 \rightarrow \cdots \rightarrow P_m \rightarrow R := fix (c_1:\sigma_1) \ldots (c_l:\sigma_l) assume h_1:P_1 \vdots assume h_m:P_m have k_1:Q_1:=\cdots \vdots have k_n:Q_n:=\cdots show R from \ldots
```

#### **4.2 Structured Constructs**

The previous section presented the main commands for writing structured proofs: fix, assume, have, and show. We now review the components of structured proofs more systematically.

#### **Theorem or Hypothesis**

The simplest structured proof, apart from sorry, is the name of a theorem or hypothesis. If we have

```
theorem two_add_two_Eq_four :
  2 + 2 = 4 :=
  by ...
```

then the theorem name two\_add\_two\_Eq\_four can be used as a proof of 2 + 2 = 4 later. For example:

```
theorem this_time_with_feeling :
  2 + 2 = 4 :=
  two_add_two_Eq_four
```

We can pass arguments to theorems to instantiate  $\forall$  quantifiers and to discharge assumptions. Suppose the theorem add\_comm (m n :  $\mathbb{N}$ ) : add m n = add n m is available, and suppose we want to prove its instance add o n = add n o. This can be achieved neatly using the name of the theorem and two arguments:

```
theorem add_comm_zero_left (n : N) :
  add ⊙ n = add n ⊙ :=
  add_comm ⊙ n
```

This is equivalent to the tactical proof by exact add\_comm o n but is more concise. The exact tactic can be seen as the inverse of by. Why enter tactic mode only to leave it immediately?

As with exact and apply, the theorem or hypothesis's statement is matched with the current goal up to computation. This gives some flexibility.

#### fix

```
fix names: type
```

The fix command moves  $\forall$ -quantified variables from the goal's target to the local context. It is a structured version of the intro tactic.

Note that the standard Lean tactic for fixing variables is fun. We prefer to use fix, which is provided by LoVelib, as a more readable alternative. We will keep fun for specifying anonymous functions.

#### assume

```
assume name: proposition
```

The assume command moves the leading assumption from goal's target to the local context. It can be seen as a structured version of the intro tactic.

Note that the standard Lean tactic for stating assumptions is fun. We prefer to use assume, which is provided by LoVelib, as a more readable alternative. We will keep fun for specifying anonymous functions.

#### have

```
have name : proposition :=
  proof
```

The have command lets us state and prove an intermediate theorem, which may refer to names introduced by previous fixes, assumes, and haves. The proof can be tactical or structured. Generally, we tend to use structured proofs to sketch the main argument and resort to tactical proofs for proving subgoals or uninteresting intermediate steps. Another kind of mixture arises when we pass arguments to theorem names. For example, given hab:  $a \rightarrow b$  and ha: a, the tactic exact hab ha will prove the goal  $\vdash$  b. Here, hab ha is a proof term nested inside a tactic.

#### let

```
let name [: type] := term
```

The let command introduces a new local definition. It can be used to name a complex object that occurs several times in the proof afterwards. It is similar to

have but is designed for computable data, not proofs. Expanding or introducing a let corresponds to  $\zeta$ -conversion (Section 3.2).

The construct let name [: type] := term must be followed by a line break or a semicolon (;).

#### show

```
show proposition from
proof
```

The show command lets us repeat the goal to prove, which can be useful as documentation. It also allows us to rephrase the goal in a syntactically equal form up to computation. Instead of the syntax show proposition from proof, we can simply write proof if we do not want to repeat the goal and do not need to rephrase it. The proof can be tactical or structured.

# 4.3 Forward Reasoning about Connectives and Quantifiers

Reasoning about the logical connectives and quantifiers in a forward fashion uses the same introduction and elimination rules as in tactic mode (Section 3.3). A few examples will give the flavor. Let us start with conjunction:

```
theorem And_swap (a b : Prop) : a \wedge b \rightarrow b \wedge a := assume hab : a \wedge b have ha : a := And.left hab have hb : b := And.right hab show b \wedge a from And.intro hb ha
```

Even readers who do not know what And.left etc. mean can understand that we extract a and b from a  $\land$  b and put them back together as b  $\land$  a. Mathematicians would probably have an easier time making sense of this proof than of its tactical counterpart:

```
theorem And_swap_tactical (a b : Prop) :
  a ∧ b → b ∧ a :=
  by
    intro hab
    apply And.intro
    apply And.right
    exact hab
    apply And.left
    exact hab
```

In general, backward proofs are easier to derive "mindlessly," and most of the automation provided by proof assistants works backwards. This makes sense: Pretend that you are Miss Marple or Hercule Poirot on a murder investigation. A backward investigation would start from the crime scene and try to extract clues that

potentially connect a handful of suspects to the crime. In contrast, a forward investigation might start by questioning as many as eight billion people to determine whether they have an alibi. Which approach is more likely to succeed?

Our next examples concern *one-point rules*. These are theorems that can be used to eliminate a quantifier when the bound variable can effectively take only one value. For example, the proposition  $\forall n$ ,  $n = 666 \rightarrow beast \geq n$  can be simplified to beast  $\geq 666$ . The following theorem justifies this simplification:

```
theorem Forall.one_point \{\alpha: \mathsf{Type}\}\ (\mathsf{t}:\alpha)\ (\mathsf{P}:\alpha\to\mathsf{Prop}): (\forall \mathsf{x},\,\mathsf{x}=\mathsf{t}\to\mathsf{P}\,\mathsf{x}) \leftrightarrow \mathsf{P}\,\mathsf{t}:= \mathsf{Iff.intro}
(\mathsf{assume hall}:\forall \mathsf{x},\,\mathsf{x}=\mathsf{t}\to\mathsf{P}\,\mathsf{x} \mathsf{show}\ \mathsf{P}\,\mathsf{t}\,\mathsf{from} \mathsf{by} \mathsf{apply}\,\mathsf{hall}\,\mathsf{t} \mathsf{rfl})
(\mathsf{assume hp}:\mathsf{P}\,\mathsf{t} \mathsf{fix}\,\mathsf{x}:\alpha \mathsf{assume heq}:\mathsf{x}=\mathsf{t} \mathsf{show}\,\mathsf{P}\,\mathsf{x}\,\mathsf{from} \mathsf{by} \mathsf{rw}\, [\mathsf{heq}] \mathsf{exact}\,\mathsf{hp})
```

The proof may look intimidating, but it was not hard to develop. The key was to proceed one step at a time. At first, we observed that the goal is an equivalence, so we wrote

```
Iff.intro (_) (_)
```

where Iff.intro is the introduction rule for  $\leftrightarrow$  (Section 3.3). The two placeholders are important to make this proof well formed. We use parentheses because we strongly suspect that the subproofs will be nontrivial. Because proofs are basically terms, which are basically programs, the advice we gave in Section 1.4 applies here as well, mutatis mutandis:

The key idea is that the proof should be syntactically correct at all times. The only red underlining we should see in Visual Studio Code should appear under the placeholders. In general, a good principle for proof development is to start with a proof that compiles, perform the smallest change possible to obtain a new compiling proof, and repeat until the proof is complete.

Hovering over the first placeholder makes the corresponding subgoal appear. We can see that Lean expects a proof of  $\vdash (\forall x, x = t \rightarrow P \, x) \rightarrow P \, t$ , so we provide a suitable skeleton. A structured proof of an implication consists of an assume followed by a show:

```
Iff.intro
  (assume hall : ∀x, x = t → P x
    show P t from
    _)
  ( )
```

Each of the remaining placeholders can be replaced by a structured proof or a tactical proof. To eliminate the first placeholder, we enter tactic mode and apply the instance  $t = t \rightarrow P$  t of the rule  $\forall x$ ,  $x = t \rightarrow P$  x to the goal  $\vdash P$  t. This leaves the proof obligation t = t, which we dischage using the rfl tactic:

```
Iff.intro
  (assume hall : \forall x, x = t \rightarrow P x
  show P t from
    by
        apply hall t
        rfl))
  (_)
```

The second placeholder stands for  $\vdash Pt \to \forall x$ ,  $x = t \to Px$ . To eliminate it, we write assume corresponding to the first implication's assumption, fix for  $\forall$ , assume for the second implication's assumption, and show for the conclusion:

```
(assume hp : P t
  fix x : α
  assume heq : x = t
  show P x from
  _)
```

The show's proof is tactical: We rewrite x to t in  $\vdash$  P x, obtaining  $\vdash$  P t, which matches the assumption hp:

```
(assume hp : P t
  fix x : α
  assume heq : x = t
  show P x from
   rw [heq]
  exact hp)
```

Let us check that the one-point rule actually works on our motivating example:

```
theorem beast_666 (beast : \mathbb{N}) : 
 (\forall n, n = 666 \rightarrow beast \ge n) \leftrightarrow beast \ge 666 := Forall.one_point _ _
```

It works. Matching Forall.one\_point t P against the statement of beast\_666 yields the instantiation t := 666 and P := (fun  $m \mapsto beast \ge m$ ).

Finally, the one-point rule for  $\exists$  demonstrates how to use the introduction and elimination rules for  $\exists$  in a structured proof:

```
theorem Exists.one_point \{\alpha: \mathsf{Type}\}\ (\mathsf{t}:\alpha)\ (\mathsf{P}:\alpha\to\mathsf{Prop}): (\exists \mathsf{x}:\alpha,\,\mathsf{x}=\mathsf{t}\land\mathsf{P}\,\mathsf{x}) \leftrightarrow \mathsf{P}\,\mathsf{t}:= \mathsf{Iff.intro}
(\mathsf{assume hex}:\exists \mathsf{x},\,\mathsf{x}=\mathsf{t}\land\mathsf{P}\,\mathsf{x} \mathsf{show}\ \mathsf{P}\,\mathsf{t}\,\mathsf{from}
\mathsf{Exists.elim\ hex}
(\mathsf{fix}\,\mathsf{x}:\alpha \mathsf{assume\ hand}:\mathsf{x}=\mathsf{t}\land\mathsf{P}\,\mathsf{x} \mathsf{have\ hxt}:\mathsf{x}=\mathsf{t}:= \mathsf{And.left\ hand}
```

```
have hpx : P x :=
    And.right hand
show P t from
    by
    rw [←hxt]
    exact hpx))
(assume hp : P t
show ∃x : α, x = t ∧ P x from
Exists.intro t
    (have tt : t = t :=
        by rfl
show t = t ∧ P t from
And.intro tt hp))
```

Although the proof might look overwhelming, it was developed mostly without thinking using the same procedure as for the proof of the one-point rule for  $\forall$ .

First, we use the introduction rule for  $\leftrightarrow$ :

```
Iff.intro ( ) ( )
```

For the first placeholder, we must provide a proof of P t under the assumption  $\exists x : \alpha, x = t \land P x$ . Thus, we expand the proof skeleton to

```
Iff.intro (assume hex : \exists x : \alpha, x = t \land P x show P t from __) (_)
```

At this point, we want to extract the witness for x in the assumption hex. To achieve this, we use the elimination rule for  $\exists$ :

```
Iff.intro
  (assume hex : ∃x : α, x = t ∧ P x
    show P t from
      Exists.elim hex (_))
  (_)
```

The first placeholder asks for a proof of the subgoal  $\vdash \forall x$ ,  $x = t \land P x \rightarrow P t$ . We sketch one:

```
Iff.intro 

(assume hex : \exists x : \alpha, x = t \land P x show P t from 

Exists.elim hex 

(fix x : \alpha assume hand : x = t \land P x show P t from __)) 

(_)
```

Notice how the proof mirrors the subgoal, with fix corresponding to  $\forall$ , assume corresponding to  $\rightarrow$ 's assumption, and show corresponding to  $\rightarrow$ 's conclusion. We can now access the witness x and its characteristic property  $x = t \land P x$ .

Next, we extract the two conjuncts of hand, using the have command and the elimination rules for  $\land$  (And.left and And.right):

```
Iff.intro
  (assume hex : ∃x : α, x = t ∧ P x
    show P t from
    Exists.elim hex
        (fix x : α
        assume hand : x = t ∧ P x
        have hxt : x = t :=
        And.left hand
        have hpx : P x :=
        And.right hand
        show P t from
        _))
  (_)
```

Now we can attempt a proof of P t. We rewrite t into x and apply the hypothesis P x:

```
Iff.intro
  (assume hex : ∃x : α, x = t ∧ P x
    show P t from
    Exists.elim hex
        (fix x : α
        assume hand : x = t ∧ P x
        have hxt : x = t :=
            And.left hand
        have hpx : P x :=
            And.right hand
        show P t from
        rw [←hxt]
        exact hpx))
  (_)
```

For the remaining placeholder, we first provide a proof skeleton based on the subgoal  $\vdash$  P t  $\rightarrow$   $\exists x: \alpha$ , x = t  $\land$  P x:

```
(assume hp : P t show \exists x : \alpha, x = t \land P x from _)
```

Then we use the introduction rule for ∃ to provide the witness t:

```
(assume hp : P t show \exists x : \alpha, x = t \land P x from Exists.intro t (_))
```

Then comes more boilerplate:

```
(assume hp : P t

show \exists x : \alpha, x = t \land P x from

Exists.intro t

(show t = t \land P t from

_))
```

The last conjunction,  $t = t \land P t$ , is easy to prove: Its left-hand side follows by reflexivity, and its right-hand side is simply the assumption hp. Finally, we get

```
(assume hp : P t
show ∃x : α, x = t ∧ P x from
Exists.intro t
   (have tt : t = t :=
        by rfl
show t = t ∧ P t from
        And.intro tt hp))
```

To develop the above structured proofs, we used largely the same procedure as for exhibiting an inhabitant of a type (Section 1.4), understanding the function arrow as implication. The similarity between the two procedures is no coincidence, as we will see in Section 4.7.

#### 4.4 Calculational Proofs

In informal mathematics, we often express proofs as transitive chains of equalities, inequalities, or equivalences (e.g., a = b = c,  $a \ge b \ge c$ , or  $a \leftrightarrow b \leftrightarrow c$ ). In Lean, such *calculational proofs* are supported by the calc command, which provides a lightweight syntax and takes care of applying transitivity theorems for preregistered relations, such as equality and the arithmetic comparison operators.

The general syntax is as follows:

```
calc
  termo op1 term1 :=
    proof1
    op2 term2 :=
    proof2
    :
    opn termn :=
    proofn
```

The underscores (\_) are part of the syntax. Each  $proof_i$  justifies the statement  $term_{i-1} \ op_i \ term_i$ . The operators  $op_i$  need not be identical, but they must be compatible with each other. For example, =, <, and  $\leq$  are compatible, whereas > and < are not.

A simple example follows:

```
theorem two_mul_example (m n : N) :
   2 * m + n = m + n + m :=
calc
   2 * m + n = m + m + n :=
   by rw [Nat.two_mul]
   _ = m + n + m :=
   by ac_rfl
```

Mathematicians (assuming they would condescend to offer a justification for such a trivial result) could have written the above proof roughly as follows:

```
2 * m + n = (m + m) + n (since 2 * m = m + m)
= m + n + m (by associativity and commutativity of +)
```

In the Lean proof, the underscore stands for the term (m + m) + n, which we would have had to repeat had we written the proof without calc:

```
theorem two_mul_example_have (m n : N) :
    2 * m + n = m + n + m :=
    have hmul : 2 * m + n = m + m + n :=
    by rw [Nat.two_mul]
    have hcomm : m + m + n = m + n + m :=
    by ac_rfl
    show _ from
    Eq.trans hmul hcomm
```

Notice that with haves, we also need to explicitly invoke Eq.trans and to give names to the two intermediate steps.

# 4.5 Forward Reasoning with Tactics

Many users prefer the tactic mode to structured proofs. But even in tactic mode, it can be useful to reason in a forward fashion, mixing forward and backward reasoning steps. The structured proof commands have, let, and calc are also available as tactics, making this possible.

The following example demonstrates the have and let tactics on a theorem we have seen several times already:

```
theorem prop_comp_tactical (a b c : Prop) (hab : a \rightarrow b) (hbc : b \rightarrow c) : a \rightarrow c := by intro ha have hb : b := hab ha let c' := c have hc : c' := hbc hb exact hc
```

#### have

```
have name : proposition :=
  proof
```

The have tactic lets us state and prove an intermediate theorem in tactic mode. Afterwards, the theorem is available as a hypothesis in the goal state.

#### let

```
let name [: type] := term
```

The let tactic lets us introduce a local definition in tactic mode. Afterwards, the defined symbol and its definition are available in the goal state.

#### calc

```
calc
  termo op1 term1 :=
    proof1
    op2 term2 :=
    proof2
    :
    opn termn :=
    proofn
```

The calc tactic lets us enter a calculational proof (Section 4.4) in tactic mode. The tactic has the same syntax as the structured proof command of the same name. We can regard calc ..., where calc is a tactic, as an alias for apply (exact calc ...), where calc is the structured proof command described above.

## 4.6 Dependent Types

Dependent types are the defining feature of the *dependent type theory* family of logics. Although you may not be familiar with the terminology, you are likely to be familiar with the concept in some form or other.

Consider a function pick that takes a natural number n (i.e., a value from  $\mathbb{N} = \{0, 1, 2, \ldots\}$ ) and that returns a natural number between 0 and n. Intuitively, pick n should have the type  $\{0, 1, \ldots, n\}$  (i.e., the type consisting of all natural numbers  $i \leq n$ ). In Lean, this is written  $\{i : \mathbb{N} // i \leq n\}$ . This would be the type of pick n. Mathematically inclined readers might want to think of pick as an  $\mathbb{N}$ -indexed family of terms

```
(pick n : \{i : \mathbb{N} // i \le n\})_{n : \mathbb{N}}
```

in which the type of each term depends on the index—e.g., pick  $5: \{i: \mathbb{N} / i \le 5\}$ . But what would be the type of pick itself? We would like to express that pick is a function that takes an argument  $n: \mathbb{N}$  and that returns a value of type  $\{i: \mathbb{N} / / i \le n\}$ . To capture this, we will write

```
pick: (n:\mathbb{N}) \to \{i:\mathbb{N} \: / / \: i \leq n\}
```

This is a dependent type: The type of the result depends on the value of the argument n. (The variable name n itself is immaterial; we could also write m or x.)

Unless otherwise specified, a *dependent type* means a type depending on a (non-type) term, as above, with  $n:\mathbb{N}$  as the term and  $\{i:\mathbb{N}\ //\ i\leq n\}$  as the type that depends on it. But:

- A type may also depend on another type—for example, the type constructor List, its  $\eta$ -expanded variant fun  $\alpha$ : Type  $\mapsto$  List  $\alpha$ , or the polymorphic type fun  $\alpha$ : Type  $\mapsto \alpha \to \alpha$  of functions with the same domain and codomain.
- A term may depend on a type—for example, the polymorphic identity function fun  $\alpha$  : Type  $\mapsto$  fun x :  $\alpha \mapsto$  x.
- And of course, a term may also depend on a term—for example, fun  $n: \mathbb{N} \mapsto n+2$ .

In summary, there are four cases for fun  $x \mapsto t$ :

Body (t)		Argument (x)	Description	
A term	depending on	a term	Simply typed λ-expression	
A type	depending on	a term	Dependent type (in the narrow sense)	
A term	depending on	a type	Polymorphic term	
A type	depending on	a type	Type constructor	

The last three rows correspond to the three axes of Henk Barendregt's  $\lambda$ -cube.<sup>1</sup>

The APP and Fun rules presented in Section 1.3 must be generalized to work with dependent types:

$$\frac{\text{C} \vdash \text{t} : (\text{x} : \sigma) \rightarrow \tau[\text{x}] \qquad \text{C} \vdash \text{u} : \sigma}{\text{C} \vdash \text{t} \text{u} : \tau[\text{u}]} \text{App'}$$

$$\frac{\text{C}, \text{x} : \sigma \vdash \text{t} : \tau[\text{x}]}{\text{C} \vdash (\text{fun } \text{x} : \sigma \mapsto \text{t}) : (\text{x} : \sigma) \rightarrow \tau[\text{x}]} \text{Fun'}$$

The notation  $\tau[x]$  stands for a type that may contain x, and  $\tau[u]$  stands for the same type where all occurrences of x have been replaced by u.

The simply typed case arises when x does not occur in  $\tau[x]$ . Then, we can simply write  $\sigma \to \text{instead}$  of  $(x:\sigma) \to .$  The familiar notation  $\sigma \to \tau$  is equivalent to  $(\_:\sigma) \to \tau$ . It is easy to check that APP' and FUN' coincide with APP and FUN when x does not occur in  $\tau[x]$ .

The example below demonstrates APP':

$$\frac{\vdash \mathsf{pick} : (\mathsf{n} : \mathbb{N}) \to \mathsf{i} : \mathbb{N} \mathbin{/\!/} \mathsf{i} \le \mathsf{n} \qquad \vdash \mathsf{5} : \mathbb{N}}{\vdash \mathsf{pick} \, \mathsf{5} : \mathsf{i} : \mathbb{N} \mathbin{/\!/} \mathsf{i} \le \mathsf{5}} \mathsf{APP}'$$

The next example demonstrates Fun':

$$\frac{\alpha: \mathsf{Type}, \mathsf{x}: \alpha \vdash \mathsf{x}: \alpha}{\alpha: \mathsf{Type} \vdash (\mathsf{fun}\, \mathsf{x}: \alpha \mapsto \mathsf{x}): \alpha \to \alpha} \mathsf{Fun}\, \mathsf{or}\, \mathsf{Fun'}}{\vdash (\mathsf{fun}\, \alpha: \mathsf{Type} \mapsto \mathsf{fun}\, \mathsf{x}: \alpha \mapsto \mathsf{x}): (\alpha: \mathsf{Type}) \to \alpha \to \alpha} \mathsf{Fun'}$$

The picture is incomplete because we only check that the terms—the entities on the left-hand side of a colon (:)—are well typed. The types—the entities on the right-hand side of a colon—should also be checked, using the same type system.

¹https://en.wikipedia.org/wiki/Lambda\_cube

For example, the type of Nat.succ is  $\mathbb{N} \to \mathbb{N}$ , whose type is Type. Types of types, such as Type and Prop, are called universes. We will study them more closely in Chapter 12.

Remarkably, universal quantification is simply an alias for a dependent type:  $\forall x: \sigma, \tau$  abbreviates  $(x:\sigma) \to \tau$ . This will become clearer below.

## 4.7 The PAT Principle

You will likely have noticed that the same symbol  $\rightarrow$  is used both for implication (e.g., False  $\rightarrow$  True) and as the type constructor of functions (e.g.,  $\mathbb{Z} \rightarrow \mathbb{N}$ ). Without context, we cannot tell whether  $a \rightarrow b$  refers to the type of a function with domain a and codomain b or to the proposition "a implies b."

It turns out that the two concepts not only *look* the same, they *are* the same. This is called the *PAT principle*, where PAT is a double mnemonic:

PAT = propositions as types PAT = proofs as terms

Furthermore, because types are also terms, it also means that propositions are terms. However, PAT is not a quadruple mnemonic (PAT = proofs as types). Also note that not all types are propositions, and not all terms are proofs.

By using terms and types to represent proofs and propositions, dependent type theory achieves a considerable economy of concepts. The question

"Is H a proof of P?"

becomes equivalent to

"Does the term H have type P?"

As a result, inside of Lean, there is no proof checker, only a type checker. Proofs are checked by the type checker.

Let us review the dramatis personae one by one. We use the mathematical variables  $\sigma$ ,  $\tau$  for types; P, Q for propositions; t, u, x for terms; and h, G, H, for proofs.

Starting with "propositions as types," for types, we have the following:

- $\sigma \to \tau$  is the type of functions from  $\sigma$  to  $\tau$ .
- $(x : \sigma) \to \tau[x]$  is the dependent type of functions from  $x : \sigma$  to  $\tau[x]$ .

In contrast, for propositions, we have the following:

- ightharpoonup P 
  ightarrow Q can be read as "P implies Q," or as the type of functions mapping proofs of P to proofs of Q.
- $\forall x : \sigma$ , Q[x] can be read as "for all x, Q[x]," or as the type of functions of type  $(x : \sigma) \rightarrow Q[x]$ , mapping values x of type  $\sigma$  to proofs of Q[x].

Continuing with "proofs as terms," for terms, we have the following:

- A constant is a term.
- A variable is a term.
- t u is the application of function t to argument u.
- fun  $x \mapsto t[x]$  is a function mapping x to t[x].

In contrast, for proofs (i.e., proof terms), we have the following:

- The name of a theorem or hypothesis is a proof.
- $\blacksquare$  H t, which instantiates the leading parameter or  $\forall$  quantifier of proof H's statement with term t, is a proof.
- H G, which discharges the leading assumption of H's statement with proof G, is a proof. This operation is called modus ponens.
- fun h :  $P \mapsto H[h]$  is a proof of  $P \to Q$ , assuming H[h] is a proof of Q for all h : P.
- fun  $x : \sigma \mapsto H[x]$  is a proof of  $\forall x : \sigma$ , Q[x], assuming H[x] is a proof of Q[x] for all  $x : \sigma$ .

The last two cases are justified by the Fun' rule. In a structured proof, as opposed to a raw proof term, we would write assume or fix instead of fun, and we would probably want to repeat the conclusion using show for readability, as follows:

```
theorem case_4: theorem case_5: P \rightarrow Q := \forall x : \sigma, \ Q[x] := assume \ h : P fix \ x : \tau show Q from H[h] H[x]
```

The terminology of dependent type theory can be quite confusing, because some words have a narrow and a broad sense. The following diagram captures the various meanings of important words:

```
term

2 + 2 : N : Type : ...

True.intro : True : Prop : ...

proposition universes

types

terms
```

In the broad sense, any expression is a term, any expression that may occur on the right-hand side of a typing judgment is a type, and any expression that may occur on the right-hand side of a typing judgment with a type on its left-hand side is a universe. This is consistent with the reading of t: u as "t has type u" and the notion that universes are types of types.

Some Lean commands have two names but are essentially the same by the PAT principle. This is the case for fix and assume; indeed, both are LoVelib aliases for fun. There are also pairs with slightly different behavior, such as def/theorem and let/have. The fundamental difference is this: When we define some function or data, we care not only about the type but also about the body—the behavior. On the other hand, once we have proved a theorem, the proof itself becomes irrelevant. All that matters is that there is a proof. We will return to the topic of proof irrelevance in Chapter 12.

The following table summarizes the differences between tactical proofs, structured proofs, and raw proof terms:

Tactical proof	Structured proof	Raw proof term
intro x	fix x : τ	$fun\;x\mapsto$
intro h	assume h : P	fun h $\mapsto$
have k := H	have k := H	(fun k $\mapsto$ ) H
<pre>let x := t</pre>	let x := t	(fun $x \mapsto$ ) t
<pre>exact (H:P)</pre>	show P from H	H : P
calc	calc	calc

Notice that let x := t followed by u is essentially a way of writing (fun  $x \mapsto u$ ) t, except that  $\beta$ -reduction is disabled.

# 4.8 Induction by Pattern Matching and Recursion

In Section 3.6, we reviewed the use of the induction tactic to perform proofs by induction. An alternative, more flexible style relies on pattern matching and the PAT principle.

Recall the definition of the reverse of a list from the end of Section 2.2.3:

```
def reverse \{\alpha : \mathsf{Type}\} : \mathsf{List} \ \alpha \to \mathsf{List} \ \alpha
| [] => []
| x :: xs => \mathsf{reverse} \ \mathsf{xs} ++ [x]
```

In fact, reverse exists as List.reverse in Lean's standard library, but our definition is designed for reasoning. A useful property to prove is that reverse is its own inverse: reverse (reverse xs) = xs for all lists xs. However, if we try to prove it by induction, we quickly run into an obstacle. The induction step is

```
ih: ∀xs, reverse (reverse xs) = xs ⊢ reverse (reverse xs ++ [x]) = x :: xs
```

Notice the unpleasant presence of ++ [x] inside the double reverse sandwich. We need a way to "distribute" the outer reverse over ++ to obtain a term that matches the induction hypothesis's left-hand side. The trick is to prove and use the following theorem:

The theorem's proof arguably looks more like a recursive function definition than a proof. The patterns on the left, [] and x :: xs, correspond to the two constructors of the  $\forall$ -quantified variable xs. On the right-hand side of each => symbol is a proof for the corresponding case. The variables on which we can pattern-match are those that appear in the  $\forall$  quantifiers, in order of occurrence (here, xs and ys). Inside the induction step's proof, the induction hypothesis is available under the same name as the theorem we are proving (reverse\_append).

We explicitly pass xs as an argument to the induction hypothesis. This restricts the hypothesis so that it only applies to xs but not to other lists. In particular, this ensures that the theorem will not be applied to x:xs, which could lead to a

circular argument: "To prove reverse\_append (x :: xs), use reverse\_append (x :: xs)." Lean's termination checker would notice that the argument is ill-founded and raise an error, but we want to avoid that. In addition, the explicit argument xs is useful as documentation, effectively saying, "The only recursive instance of the theorem we need is the one on xs."

For reference, a tactical proof would be as follows:

The theorem would also be provable, and useful, if we put [y] instead of ys. But it is a good habit to state theorems as generally as possible. This results in more reusable libraries. Moreover, this is often necessary to obtain a strong enough induction hypothesis in a proof by induction. In general, finding the right inductions and theorems can require thought and creativity.

Simultaneous pattern matching on multiple variables is supported (e.g., xs and ys above). The patterns are then separated by commas. The general format is

Notice the strong similarity with the syntax of def (Section 2.2). The two commands are, in fact, almost the same, but theorem considers the defined term or the proof opaque, whereas def keeps it transparent. Since the actual proofs are irrelevant once a theorem is proved (Section 12), there is no need to expand them later. A similar distinction exists between let and have.

By the PAT principle, a proof by induction by pattern matching and recursion is the same as a recursive proof term. When we invoke the induction hypothesis, we are really just invoking a recursive function recursively. This explains why the induction hypothesis has the same name as the theorem we prove. Lean's termination checker is used to establish well-foundedness of the proof by induction.

With the reverse\_append theorem in place, we can return to our initial goal:

```
theorem reverse_reverse {α : Type} :
    ∀xs : List α, reverse (reverse xs) = xs
    | [] => by rfl
    | x :: xs =>
    by simp [reverse, reverse_append, reverse_reverse xs]
```

Induction by pattern matching and recursion is popular among Lean users. Its main advantages are its convenient syntax and its support for well-founded induction, which is more powerful than structural induction as provided by the induction tactic (Section 3.7). However, in this guide, we will not need the full

power of well-founded induction. Furthermore, for subtle logical reasons, induction by pattern matching and recursion is not available for inductive predicates, which are the topic of Chapter 6. For these reasons, we will often prefer the induction tactic.

# 4.9 Summary of Newly Introduced Lean Constructs

#### **Proof Commands**

assume states assumptions
calc combines proofs by transitivity
fix fixes variables
have states an intermediate theorem
let introduces a local definition
show states the target

#### **Tactics**

calc combines proofs by transitivity states an intermediate theorem let introduces a local definition

# Part II Functional-Logic Programming

# **Chapter 5**

# **Functional Programming**

We take a closer look at the essence of typed functional programming: inductive types, proofs by induction, recursive functions, pattern matching, structures (records), and type classes. The concepts covered here are described in more detail in Chapters 7 to 10 of *Theorem Proving in Lean 4* [13].

# **5.1 Inductive Types**

Inductive types are modeled after the data types of typed functional programming languages (e.g., Haskell, ML, OCaml). They are also reminiscent of sealed classes in Scala. In Chapter 2, we saw some basic inductive types: the natural numbers, a type of arithmetic expressions, and the finite lists. In this chapter, we revisit the lists and study binary trees. We also take a brief look at vectors of length n, a dependent type.

Recall the definition of natural numbers as an inductive type:

```
inductive Nat : Type where
   | zero : Nat
   | succ : Nat → Nat
```

This defines the type Nat and the two constants Nat.zero and Nat.succ, called constructors. The definition also asserts some properties of the constructors. In addition, it introduces further constants that are used internally to support induction and recursion.

As we saw in Section 2.1, an inductive type is a type whose members are all the values that can be built by a finite number of applications of its constructors, and only those. Mottos:

- No junk: The type contains no values beyond those expressible using the constructors.
- *No confusion*: Values built using a different combination of constructors are different.

For natural numbers, "no junk" means that there are no special values (such as -1,  $\varepsilon$ ,  $\infty$ , or NaN) that cannot be expressed using a finite combination of Nat.zero and Nat.succ, and "no confusion" ensures that Nat.zero  $\neq$  Nat.succ n for all n and that Nat.succ is injective. In addition, values of inductive types are always

finite. The infinite term

```
Nat.succ (Nat.succ (Nat.succ (Nat.succ ...)))
```

is not a value. Nor does there exist a value n such that Nat.succ n = n, as we will show below.

Inductive types are very convenient to use, because they support induction and recursion and their constructors are well behaved, but not all types can be defined as an inductive type. In particular, mathematical types such as  $\mathbb Q$  (the rationals) and  $\mathbb R$  (the real numbers) require more elaborate constructions, based on quotienting and subtyping. This will be explained in Chapters 12 and 14.

## **5.2 Structural Induction**

Structural induction is a generalization of mathematical induction to arbitrary inductive types. To prove a goal  $n: \mathbb{N} \vdash P[n]$  by structural induction on n, it suffices to show two subgoals, traditionally called the base case and the induction step:

```
⊢ P[0]
k: N,ih: P[k] ⊢ P[k+1]
```

We can of course also write P[Nat.zero] and P[Nat.succ k].

In general, the situation is more complex. The goal might contain some extra hypotheses (e.g., Q) that do not depend on n and others (e.g., R[n]) that do. Assuming we have one hypothesis of each kind, this gives the initial goal

```
hQ:Q, n:\mathbb{N}, hR:R[n] \vdash S[n]
```

Structural induction on n then produces the two subgoals

```
\begin{array}{c} \mathsf{hQ}: \mathsf{Q}, \, \mathsf{hR}: \, \mathsf{R}[\mathfrak{0}] \, \vdash \, \mathsf{S}[\mathfrak{0}] \\ \mathsf{hQ}: \mathsf{Q}, \, \mathsf{k}: \, \mathbb{N}, \, \mathsf{ih}: \, \mathsf{R}[\mathsf{k}] \rightarrow \mathsf{S}[\mathsf{k}], \, \mathsf{hR}: \, \mathsf{R}[\mathsf{k}+1] \, \vdash \, \mathsf{S}[\mathsf{k}+1] \end{array}
```

The hypothesis Q is simply carried over unchanged from the initial goal, whereas  $R[n] \vdash S[n]$  is treated almost the same as if the goal's target had been  $R[n] \rightarrow S[n]$ . This is easy to check by taking  $P[n] := R[n] \rightarrow S[n]$  in the first example above. Since this general format is very verbose and hardly informative (now that we understand how it works), from now on we will present goals in the simplest form possible, without extra hypotheses.

For lists, given a goal xs : list  $\alpha \vdash P[xs]$ , structural induction on xs yields

```
\vdash P[[]]
y: \alpha, ys: list \alpha, ih: P[ys] \vdash P[y::ys]
```

We can of course also write List.nil and List.cons y ys. There is no induction hypothesis associated with y, because y is not of list type.

For arithmetic expressions, the base cases are

```
i: ℤ ⊢ P[AExp.numi] x: String ⊢ P[AExp.varx]
```

and the induction steps are

```
e_1 e_2 : AExp, ih_1 : P[e_1], ih_2 : P[e_2] \vdash P[AExp.add e_1 e_2]
e_1 e_2 : AExp, ih_1 : P[e_1], ih_2 : P[e_2] \vdash P[AExp.sub e_1 e_2]
e_1 e_2 : AExp, ih_1 : P[e_1], ih_2 : P[e_2] \vdash P[AExp.mul e_1 e_2]
e_1 e_2 : AExp, ih_1 : P[e_1], ih_2 : P[e_2] \vdash P[AExp.div e_1 e_2]
```

Notice the two induction hypotheses, about  $e_1$  and  $e_2$ .

In general, structural induction produces one subgoal per constructor. In each subgoal, induction hypotheses are available for all constructor arguments of the type we are performing the induction on.

Given an inductive type  $\tau$ , the procedure to compute the subgoals is always the same:

- 1. Replace the hole in P[] with each possible constructor applied to fresh variables (e.g., y:: ys), yielding as many subgoals as there are constructors.
- 2. Add these new variables (e.g., y, ys) to the local context.
- 3. Add induction hypotheses for all new variables of type  $\tau$ .

As an example, we will prove that Nat.succ  $n \neq n$  for all  $n : \mathbb{N}$ . We start with an informal proof:

The proof is by structural induction on n.

CASE 0: We must show Nat.succ  $0 \neq 0$ . This follows from the "no confusion" property of the constructors of inductive types.

CASE Nat.succ k: The induction hypothesis is Nat.succ  $k \neq k$ . We must show Nat.succ (Nat.succ k)  $\neq$  Nat.succ k. By the injectivity of Nat.succ, we have that Nat.succ (Nat.succ k) = Nat.succ k is equivalent to Nat.succ k = k. Thus, it suffices to prove Nat.succ  $k \neq k$ , which corresponds exactly to the induction hypothesis.

Notice the main features of this informal proof, which you should aim to reproduce in your own informal arguments:

- The proof starts with an unambiguous announcement of the type of proof we are carrying out (e.g., which kind of induction and on which variable).
- The cases are clearly identified, and for each case, both the goal's target and the hypotheses are stated.
- The key theorems on which the proof relies are explicitly invoked (e.g., injectivity of Nat.succ).

Now let us carry out the proof in Lean:

The routine reasoning about constructors is all carried out by simp.

## **5.3 Structural Recursion**

Structural recursion is a form of recursion that allows us to peel off one constructor from the value on which we recurse. The factorial function below is structurally recursive:

```
def fact : \mathbb{N} \to \mathbb{N}

\mid 0 => 0

\mid n + 1 => (n + 1) * fact n
```

The constructor we peel off here is Nat.succ (written + 1). Such functions are guaranteed to call themselves only finitely many times before the recursion stops; for example, fact 12345 will call itself 12345 times. The function is said to *terminate*. This property helps ensure logical consistency.

With structural recursion, there are as many equations as there are constructors. Novices are often tempted to supply additional, redundant cases, as in the following example:

```
def factThreeCases : \mathbb{N} \to \mathbb{N}

| 0 => 0

| 1 => 1

| n + 1 => (n + 1) * factThreeCases n
```

It is in your own best interest to resist this temptation. The more cases you have in your definitions, the more work it will be to reason about them. Keep in mind the saying that one good definition is worth three theorems.

For structurally recursive functions, Lean can automatically prove termination. For more general recursive schemes, the termination check may fail. Sometimes it does so for a good reason, as in the following example:

```
-- fails  \begin{tabular}{ll} \begin{tabular
```

If Lean were to accept this definition, we could exploit it to prove that o = 1, by subtracting illegal n from each side of the equation illegal n = illegal n + 1. From o = 1, we could derive False, and from False, we could derive anything. Clearly, we do not want that.

If we had used opaque and axiom, nothing could have saved us:

```
opaque immoral : N → N

axiom immoral_eq (n : N) :
   immoral n = immoral n + 1

theorem proof_of_False :
   False :=
   have hi : immoral 0 = immoral 0 + 1 :=
      immoral_eq 0
   have him :
      immoral 0 - immoral 0 = immoral 0 + 1 - immoral 0 :=
      by rw [←hi]
   have hoeq1 : 0 = 1 :=
```

```
by simp at him
show False from
by simp at hoeq1
```

Another reason for preferring def over opaque and axiom is that the defining equations can be used in computations. Tactics such as rfl that unify up to computation become stronger each time we introduce a definition, and the diagnostic commands #eval and #reduce can be used on defined constants.

The observant reader will have noticed that the above definitions of factorial are mathematically wrong: fact shockingly returns o regardless of the argument. We quite literally facted up. These embarrassing mistakes remind us to *test* our definitions and *prove* some of their properties. Although flawed axioms arise now and then, what is much more common are definitions that fail to capture the intended concepts. Just because a function is called fact does not mean that it actually computes factorials.

# 5.4 Pattern Matching Expressions

Pattern matching is possible not only at the top level of a def command but also deeply within terms, via a match expression. The construct has the following general syntax:

```
match term1, ..., termm with
  | pattern11, ..., pattern1m => result1
    :
  | pattern11, ..., patternnm => resultn
```

The construct is reminiscent of match in some modern programming languages. An English translation of the above match expression follows:

```
Consider the terms term_1, ..., term_m.

If they are respectively of the forms pattern_{11}, ..., pattern_{1m}, then yield result_1.

\vdots

If they are respectively of the forms pattern_{n1}, ..., pattern_{nm}, then yield result_n.
```

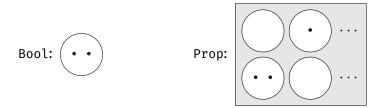
The patterns may contain variables, constructors, and nameless placeholders (\_). The  $result_i$  expressions may refer to the variables introduced in the corresponding patterns.

The following function definition demonstrates the syntax of pattern matching within expressions:

```
\begin{array}{lll} \text{def bcount } \{\alpha: \text{Type}\} \; (\text{p}: \alpha \to \text{Bool}): \text{List } \alpha \to \mathbb{N} \\ & \mid [] & => 0 \\ & \mid x:: xs => \\ & \text{match p x with} \\ & \mid \text{true} => \text{bcount p xs} + 1 \\ & \mid \text{false} => \text{bcount p xs} \end{array}
```

The bcount function counts the number of elements in a list that satisfy the given predicate p. The predicate's codomain is Bool. As a general rule, we will use type Bool, of Booleans, within programs and use the type Prop, of propositions, when stating properties of programs. The two values of type Bool are called false and true (in lowercase). The connectives are called or (infix: ||), and (infix: &&), and not (prefix: !).

The following diagram shows the interpretations of Bool and Prop:



Dots represents elements, circles represents types, and the rectangle represents a type of types, or universe. We see that Bool is interpreted as a set with two values, whereas Prop consists of an infinite number of propositions (the types), each of which has zero or more proofs (the elements). We will refine this picture in Chapter 6.

We cannot match on a proposition (of type Prop), but we can use if-thenelse instead. For example, the min operator on natural numbers operator can be defined as follows:

```
def min (a b : \mathbb{N}) : \mathbb{N} := if a \leq b then a else b
```

This requires a decidable (i.e., executable) proposition. This is the case for  $\leq$ : Given concrete values for the arguments, such as 35 and 49, Lean can reduce 35  $\leq$  49 to True. Lean keeps track of decidability using a mechanism called type classes, which will be explained below.

#### **5.5 Structures**

Lean provides a convenient syntax for defining *structures* (also called records). These are essentially nonrecursive single-constructor inductive types but with some syntactic sugar.

The definition below introduces a structure called RGB with three fields of type  $\mathbb N$  called red, green, and blue:

```
\begin{array}{cccc} \textbf{structure} & \textbf{RGB} & \textbf{where} \\ & \textbf{red} & : & \mathbb{N} \\ & \textbf{green} & : & \mathbb{N} \\ & \textbf{blue} & : & \mathbb{N} \end{array}
```

This definition has roughly the same effect as the following commands:

```
inductive RGB : Type where |\ \mathsf{mk}\ :\ \mathbb{N}\ \to\ \mathbb{N}\ \to\ \mathbb{R}\ \to\ \mathsf{RGB}
```

<sup>&</sup>lt;sup>1</sup>Lean also allows us to use True and False, but these are then implicitly converted from Prop to Bool. We generally recommend avoiding such implicit coercions. Unfortunately, there is no way to disable them.

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```
\begin{array}{l} \text{def RGB.red} : \text{RGB} \to \mathbb{N} \\ \mid \text{RGB.mk r} \_ => \text{r} \\ \\ \text{def RGB.green} : \text{RGB} \to \mathbb{N} \\ \mid \text{RGB.mk} \_ \text{g} \_ => \text{g} \\ \\ \text{def RGB.blue} : \text{RGB} \to \mathbb{N} \\ \mid \text{RGB.mk} \_ \_ \text{b} => \text{b} \end{array}
```

We can define a new structure as the extension of an existing structure. The definition below extends RGB with a fourth field, called alpha:

```
structure RGBA extends RGB where alpha : \mathbb{N}
```

The general syntax to define structures is

The parameters  $params_1$ , ...,  $params_k$  are effectively additional fields, but unlike field- $name_1$ , ..., field- $name_n$ , they are stored in the type, as arguments to the type constructor (structure-name).

Values can be specified in a variety of syntaxes:

The definition of semitransparentGreen copies all the values from pureGreen except for the alpha field, which it sets explicitly.

Next, we define an operation called shuffle:

```
def shuffle (c : RGB) : RGB :=
  { red := RGB.green c
   green := RGB.blue c
  blue := RGB.red c }
```

The definition relies on the generated selectors RGB.red, RGB.green, and RGB.blue. Instead of RGB.red c, we could also write c.red, and similarly for the other fields. Sometimes we will see this notation in Lean's output, even if we do not use it ourselves.

Applying shuffle three times in a row is the same as not applying it at all:

```
theorem shuffle_shuffle (c : RGB) :
    shuffle (shuffle c)) = c :=
    by rfl
```

### 5.6 Type Classes

Type classes are a mechanism that was popularized by Haskell and that is present in several proof assistants. In Lean, a *type class* is a structure type combining abstract constants and their properties.<sup>2</sup> A type can be declared an instance of a type class by providing concrete definitions for the constants and proving that the properties hold. Based on the type, Lean retrieves the relevant instance.

A simple example is the type class Inhabited, which requires only a constant Inhabited.default and no properties:

```
class Inhabited (\alpha : Type) : Type where default : \alpha
```

The syntax is the same as for structures, except with the keyword class instead of structure. The parameter  $\alpha$  represents an arbitrary type that could be a member of this class. This particular type class has a single parameter and a single field, but in general a type class can have multiple parameters and multiple fields.

Any type that has at least one element can be registered as an instance of the Inhabited type class. For example, we can register  $\mathbb N$  by choosing an arbitrary number to be the default value:

```
instance Nat.Inhabited : Inhabited N :=
  { default := 0 }
```

This specifies a value called Nat.Inhabited of type Inhabited  $\mathbb{N}$ . Because we use the keyword instance instead of def, the structure value is registered as *the* canonical instance to use whenever a structure of type Inhabited  $\mathbb{N}$  is desired. In the global table storing type class instances, there is now an entry

```
Inhabited \mathbb{N} \mapsto \mathsf{Nat}.\mathsf{Inhabited}
```

For lists, the empty list is an obvious default value that can be constructed even if  $\alpha$  is not inhabited:

```
instance List.Inhabited \{\alpha : \mathsf{Type}\} : \mathsf{Inhabited} \ (\mathsf{List} \ \alpha) := \{ \mathsf{default} := [] \}
```

This adds the following entry to the global table:

```
Inhabited (List ?\alpha) \mapsto List.Inhabited
```

Sometimes we may want to supply several instances of a given type class for the same type. Lean will then choose the first matching instance it finds.

As an example, observe that finite functions of type  $\alpha \to \beta$  can be represented by their function tables, of type  $\beta \times \cdots \times \beta$  (with  $|\alpha|$  copies of  $\beta$ ). Accordingly,

<sup>&</sup>lt;sup>2</sup>Despite its name, Lean's type class mechanism is more closely related to Scala's implicit arguments than to Haskell's type classes.

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 $|\alpha \to \beta| = |\beta|^{|\alpha|}$ . To make this o, we must have both  $|\beta| = o$  and  $|\alpha| \ne o$ . In other words, the type  $\alpha \to \beta$  is inhabited if either (1)  $\beta$  is inhabited or (2)  $\alpha$  is *not* inhabited. We focus on case (1):

```
instance Fun.Inhabited \{\alpha \ \beta : \text{Type}\}\ [\text{Inhabited}\ \beta] :
Inhabited (\alpha \to \beta) :=
\{ \text{ default } := \text{ fun } a : \alpha \mapsto \text{Inhabited.default } \}
```

The instance relies itself on an instance of the same type class but on a different type. This occurs frequently.

The type  $\alpha \times \beta$  of pairs, also called the product type, contains values of the form (a, b), where a:  $\alpha$  and b:  $\beta$ . Given a pair ab:  $\alpha \times \beta$ , the first and second components can be extracted by writing Prod.fst ab and Prod.snd ab. To provide an inhabitant of  $\alpha \times \beta$ , we need both an inhabitant of  $\alpha$  and an inhabitant of  $\beta$ :

```
instance Prod.Inhabited \{\alpha \ \beta : \ Type\}
[Inhabited \alpha] [Inhabited \beta]:
Inhabited (\alpha \times \beta) :=
\{ \ default := (Inhabited.default, Inhabited.default) \}
```

Using the Inhabited type class, we can define the head operation on lists: the function that returns the first element of a list. Because an empty list contains no elements, there is no meaningful value we can return in that case. Given a type that belongs to the Inhabited type class, we can simply return the default value:

```
def head \{\alpha: \mathsf{Type}\} [Inhabited \alpha] : List \alpha \to \alpha | [] => Inhabited.default | x :: _ => x
```

We require that  $\alpha$  belongs to Inhabited by writing [Inhabited  $\alpha$ ]. This allows us to access Inhabited.default in the definition.

The syntax [Inhabited  $\alpha$ ] adds an implicit argument of the head constant. But unlike for other implicit arguments, Lean performs a type class search through all declared instances to determine the value of this argument. Thus, when running the command

```
#eval head ([] : List \mathbb{N})
```

Lean will look for an Inhabited  $\mathbb N$  instance and find Nat.Inhabited, the instance we declared above. In that declaration, we set default to be  $\mathfrak o$  and hence this is what #eval prints. If multiple instances are applicable and Lean chooses the wrong one, we can use the  $\mathfrak d$  syntax to transform type class arguments into explicit arguments and supply the desired type class instance.

Let us take a closer look at Inhabited.default:

```
Inhabited.default \{\alpha : \mathsf{Type}\}\ [\mathsf{Inhabited}\ \alpha] : \alpha
```

Notice that the selectors of plain structures use parentheses ( ) whereas those of type classes use square brackets [ ]. When we used Inhabited.default  $\alpha$  to define head, Lean looked for an instance of Inhabited  $\alpha$  in the global table of registered instances and in the local context. The global table contained an entry for Inhabited  $\mathbb N$  but none that matched Inhabited  $\alpha$ . On the other hand, the local context contained an anonymous parameter of type Inhabited  $\alpha$ , which could be used.

Lean's core library defines List.head exactly as we did. In practice, almost all types are nonempty (with the notable exception of False), so the Inhabited restriction is hardly an issue.

We can prove abstract theorems about the Inhabited type class, such as

```
theorem head_head \{\alpha : \mathsf{Type}\}\ [\mathsf{Inhabited}\ \alpha]\ (\mathsf{xs} : \mathsf{List}\ \alpha) : head [\mathsf{head}\ \mathsf{xs}] = \mathsf{head}\ \mathsf{xs}
```

The assumption [Inhabited  $\alpha$ ] is needed to use the operator head on lists of type List  $\alpha$ . If we omit this assumption, Lean will raise an error telling us that type class synthesis failed.

There are more type classes with only a constant but no properties, including

```
class Zero (\alpha : Type) where class One (\alpha : Type) where one : \alpha class Neg (\alpha : Type) where neg : \alpha \to \alpha class Add (\alpha : Type) where inv : \alpha \to \alpha class Mul (\alpha : Type) where add : \alpha \to \alpha \to \alpha class Mul (\alpha : Type) where mul : \alpha \to \alpha \to \alpha
```

These syntactic type classes introduce constants that are used in different contexts with different semantics. For example, one can stand for the natural number 1, the integer 1, the real 1, the identity matrix, and many other concepts of oneness. The main purpose of these type classes is to form the foundation for a rich hierarchy of algebraic type classes (groups, monoids, rings, fields, etc.) and to allow the overloading of common mathematical symbols such as +, \*, 0, 1, and  $^{-1}$ 

Syntactic type classes do not impose severe restrictions on the types that can be declared instances. In contrast, the *semantic* type classes contain properties that restrict how the given constants behave.

In Section 3.6, we encountered the following semantic type classes:

```
class IsCommutative (\alpha: Type) (f: \alpha \to \alpha \to \alpha) where comm: \forall a b, f a b = f b a class IsAssociative (\alpha: Type) (f: \alpha \to \alpha \to \alpha) where assoc: \forall a b c, f (f a b) c = f a (f b c)
```

This time, the associations are not from a type to a constant, but from a type and a function to a property. Lean does not mind the abuse: Although they are called type classes, Lean's type classes are very flexible and can be used to express all kinds of constraints.

Conceptually, IsCommutative is a dependent type of triples ( $\alpha$ , f, comm), and similarly for IsAssociative. The type of f depends on  $\alpha$ , and the type of comm depends on  $\alpha$  and f. Although they are parameters,  $\alpha$  and f are also stored along with comm.

In Section 3.6, we registered our add function on  $\mathbb N$  as a commutative and associative operation:

```
instance IsAssociative_add : IsAssociative N add :=
```

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```
{ assoc := add_assoc }
instance IsCommutative_add : IsCommutative N add :=
  { comm := add_comm }
```

Whenever we try to access @IsCommutative.comm  $\mathbb N$  add, we obtain add\_comm, and similarly for @IsAssociative.assoc  $\mathbb N$  add. The ac\_rfl tactic tries to look up the comm and assoc properties for all binary operators in the problem and exploits the properties whenever they are present.

The general syntax to define a type class is as follows:

```
class class-name (params<sub>1</sub>: type<sub>1</sub>) ... (params<sub>k</sub>: type<sub>k</sub>)
  [extends structure<sub>1</sub>, ..., structure<sub>m</sub>] where
  constant-name<sub>1</sub>: constant-type<sub>1</sub>
    :
  constant-name<sub>n</sub>: constant-type<sub>n</sub>
  property-name<sub>1</sub>: proposition<sub>1</sub>
    :
  property-name<sub>p</sub>: proposition<sub>p</sub>
```

The general syntax to instantiate a type class is as follows:

## 5.7 Lists

Lean provides a rich library of functions on finite lists. In this section, we will review some of them, and we will define some of our own; these are good exercises to familiarize ourselves with functional programming in Lean.

In the first example, we perform a case distinction on a list:

The proof relies on cases, a relative of induction. It performs a case distinction on its argument but does not generate induction hypotheses. The invocation cases xs transforms a goal  $\vdash P[xs]$  into two subgoals,  $\vdash P[[]]$  and  $\vdash P[x::xs']$ . We could also have used induction xs. If you find yourself hesitating between induction and cases, you can always choose induction and see afterwards if

you needed the induction hypotheses—if you did not need it, it is good style to document that fact by replacing induction by cases.

In a structured proof, we can use match expressions to perform a case distinction:

```
theorem head_head_match \{\alpha: Type\} [Inhabited \alpha] 
 (xs : List \alpha) : head [head xs] = head xs := match xs with 
 | List.nil => by rfl 
 | List.cons x xs' => by rfl
```

In the next example, we show how to exploit injectivity of constructors. The cases tactic can be used to exploit injectivity to simplify equations in which both sides have the same constructor applied. In the proof below, the equation before simplification is x :: xs = y :: ys:

```
theorem injection_example \{\alpha: \mathsf{Type}\}\ (x\ y:\alpha)\ (xs\ ys: \mathsf{List}\ \alpha)   (h:x::xs=y::ys):   x=y\wedge xs=ys:=   by   cases h  simp
```

The cases tactic replaces y by x and ys by xs throughout the goal, yielding the subgoal

```
\vdash x = x \land xs = xs
```

which simp can easily prove.

The cases tactic is also useful when the constructors are different, to detect the impossible case:

```
theorem distinctness_example \{\alpha : \mathsf{Type}\}\ (y : \alpha)\ (ys : \mathsf{List}\ \alpha) (h : [] = y :: ys) : false := by cases h
```

Next, we define the *map function* on lists: a function that applies its argument f—which is itself a function—to all elements stored in a collection.

```
def map \{\alpha \ \beta : \mathsf{Type}\}\ (\mathsf{f} : \alpha \to \beta) : \mathsf{List}\ \alpha \to \mathsf{List}\ \beta | [] => [] | x :: xs => f x :: map f xs
```

Notice that because f does not change in the recursive call, we put it in as a parameter of the entire definition. The alternative, which is the only option for arguments that change in recursive calls, would be as follows:

```
def mapArgs \{\alpha \ \beta : \mathsf{Type}\} : (\alpha \to \beta) \to \mathsf{List} \ \alpha \to \mathsf{List} \ \beta  | _, [] => [] | f, x :: xs => f x :: mapArgs f xs
```

A basic property of map functions is that they have no effect if their argument is the identity function (fun  $x \mapsto x$ ):

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```
theorem map_ident \{\alpha : \mathsf{Type}\}\ (\mathsf{xs} : \mathsf{List}\ \alpha) :
map (\mathsf{fun}\ \mathsf{x} \mapsto \mathsf{x})\ \mathsf{xs} = \mathsf{xs} :=
by
induction \mathsf{xs} with
|\ \mathsf{nil}\ |\ \mathsf{simp}\ [\mathsf{map},\ \mathsf{ih}]
```

Another basic property is that successive maps can be compressed into a single map, whose argument is the composition of the functions involved:

```
theorem map_comp {\alpha \beta \gamma : Type} (f : \alpha \to \beta) (g : \beta \to \gamma) (xs : List \alpha) : map g (map f xs) = map (fun x \mapsto g (f x)) xs := by induction xs with | nil => rfl | cons x xs' ih => simp [map, ih]
```

When introducing new operations, it is useful to show how these behave when used in combination with other operations. Here is an example:

```
theorem map_append \{\alpha \ \beta : \text{Type}\}\ (f: \alpha \to \beta) (xs ys : List \alpha) : map f (xs ++ ys) = map f xs ++ map f ys := by induction xs with | nil => rfl | cons x xs' ih => simp [map, ih]
```

Remarkably, the last three proofs are textually identical. These are typical induction-rfl-simp proofs.

The next list operation removes the first element of a list, returning the tail:

```
def tail \{\alpha : \mathsf{Type}\} : \mathsf{List}\ \alpha \to \mathsf{List}\ \alpha | [] => [] | _ :: xs => xs
```

For [], we simply return [] as its own tail.

The counterpart of tail is a function that extracts the first element of a list. We already reviewed one solution in Section 5.6, using the Inhabited type class. Another possible definition uses an Option wrapper:

```
def headOpt \{\alpha : \mathsf{Type}\} : \mathsf{List}\ \alpha \to \mathsf{Option}\ \alpha | [] => Option.none | x :: => Option.some x
```

The type Option  $\alpha$  has two constructors: Option.none and Option.some a, where a:  $\alpha$ . We use Option.none when we have no meaningful value to return and Option.some otherwise. We can think of Option.none as the null pointer of functional programming, but unlike null pointers (and null references), the type system guards against unsafe dereferences. To retrieve the value stored in an Option, we must pattern-match. Schematically:

```
match headOpt xs with
   | Option.none => handleTheError
   | Option.some x => doSomethingWithValue x
```

We cannot simply write doSomethingWithValue (headOpt xs), because this would be type-incorrect. The type system forces us to think about error handling.

Using the power of dependent types, another way to implement a partial function is to specify a precondition. The caller must then pass a proof that the precondition is satisfied as argument:

```
def headPre \{\alpha: \mathsf{Type}\}: (\mathsf{xs}: \mathsf{List}\ \alpha) \to \mathsf{xs} \neq [] \to \alpha | [], hxs => by simp at * | x :: _, hxs => x
```

The headPre function takes two explicit arguments. The first argument, xs, is a list. The second argument, hxs, is a proof of xs  $\neq$  []. Since the type xs  $\neq$  [] of the second argument depends on the first argument, we must use the dependent type syntax (xs: List  $\alpha$ )  $\rightarrow$  rather than List  $\alpha$   $\rightarrow$  so that we can name the first argument. The result of the function is a value of type  $\alpha$ ; thanks to the precondition, there is no need for an Option wrapper.

The second argument is used to rule out the case where xs is []. In that case, the argument (called hxs) is a proof of []  $\neq$  [], which is impossible. The proof derives a contradiction and exploits it to derive an arbitrary  $\alpha$ . From a contradiction, we can derive anything, even an inhabitant of  $\alpha$ .

We can then invoke the function as follows:

```
#eval headPre [3, 1, 4] (by simp)
```

This prints 3. The second argument, by simp, is a proof that [3, 1, 4] is not []. Now, given two lists  $[x_1, ..., x_n]$  and  $[y_1, ..., y_n]$  of the same length, the zip operation constructs a list of pairs  $[(x_1, y_1), ..., (x_n, y_n)]$ :

The function is also defined if one list is shorter than the other. For example, zip[a, b, c][x, y] = [(a, x), (b, y)]. Notice that the recursion, with three cases, deviates slightly from the structural recursion schema.

The length of a list is defined by recursion:

```
def length \{\alpha : \mathsf{Type}\} : \mathsf{List} \ \alpha \to \mathbb{N}

| \ [ \ ] => 0

| \ \mathsf{x} :: \mathsf{xs} => \mathsf{length} \ \mathsf{xs} + \mathsf{1}
```

We can say something interesting about the length of zip's result—namely, it is the minimum of the lengths of the two input lists:

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The proof above teaches us yet another trick. The induction hypothesis is

```
ih: \forallys: list \beta, length (zip xs ys) = min (length xs) (length ys)
```

Why is there a  $\forall$  quantifier? The induction xs generalizing ys tactic generalized the theorem statement so that the induction hypothesis is not restricted to some fixed ys as the proof goal but can be used for arbitrary values of ys. Such flexibility is needed here because we want to instantiate the quantifier with ys's tail (called ys') and not with ys itself.

The proof relies on a theorem about the min function that we need to prove ourselves:

```
theorem min_add_add (l m n : N) :
    min (m + l) (n + l) = min m n + l :=
    by
        cases Classical.em (m ≤ n) with
        | inl h => simp [min, h]
        | inr h => simp [min, h]
```

Recall the definition min a b = (if a  $\leq$  b then a else b). To reason about min, we often need to perform a case distinction on the condition a  $\leq$  b. This is achieved using cases Classical.em (a  $\leq$  b). This creates two subgoals: one with a  $\leq$  b as a hypothesis and one with  $\neg$  a  $\leq$  b.

Here are two different ways to perform a case distinction on a proposition in a structured proof:

```
theorem min_add_add_match (l m n : N):
    min (m + l) (n + l) = min m n + l :=
    match Classical.em (m \le n) with
    | Or.inl h => by simp [min, h]
    | Or.inr h => by simp [min, h]

theorem min_add_add_if (l m n : N):
    min (m + l) (n + l) = min m n + l :=
    if h : m \le n then
        by simp [min, h]
    else
        by simp [min, h]
```

We see again that the mechanisms that are available to write functional programs, such as match and if-then-else, are also available for writing structured proofs (which are, after all, terms). We can now add a few rows to the table presented at the end of Section 4.7:

Tactical proof	Structured proof	Raw proof term
cases t cases Classical.em Q	match t with if Q then else	match t with if Q then else

We conclude with a distributivity law about map and zip, expressed using the Prod.fst and Prod.snd selectors on pairs:

```
theorem map_zip \{\alpha \ \alpha' \ \beta \ \beta' : \text{Type}\}\ (f:\alpha \to \alpha') (g:\beta \to \beta'): \forall xs \ ys, map (fun ab : \alpha \times \beta \mapsto (f (Prod.fst ab), g (Prod.snd ab))) (\text{zip } xs \ ys) = \text{zip } (\text{map } f \ xs) \ (\text{map } g \ ys) | \ x :: xs, \ y :: ys => by \ \text{simp } [\text{zip, map, map_zip } f \ g \ xs \ ys] | \ [], \qquad => by \ \text{rfl} | \ =: \ \_, \ [] \qquad => by \ \text{rfl}
```

The patterns on the left correspond exactly to the patterns used in the definition of zip. This is simpler than performing the induction on xs and the case distinction on ys separately, as we did when we proved length\_zip. Good proofs often follow the structure of the definitions they are based on.

In the definition of zip and in the proof of map\_zip, we were careful to specify three nonoverlapping patterns. It is also possible to write equations with overlapping patterns, as in

```
\begin{array}{lll} \text{def f} & \{\alpha : \text{Type}\} : \text{List } \alpha \to \cdots \\ & | & [] => \ldots \\ & | & \text{xs =>} \ldots \text{ xs } \ldots \end{array}
```

Since the patterns are applied sequentially, the above command defines the same function as

```
def f {\alpha : Type} : List \alpha \rightarrow \cdots | [] => ... | x :: xs => ... (x :: xs) ...
```

We generally recommend the latter, more explicit style, because it leads to fewer surprises.

# 5.8 Binary Trees

Inductive types with constructors taking several recursive arguments define treelike objects. *Binary trees* have nodes with at most two children. A possible definition of binary trees follows:

```
inductive BinTree (\alpha : Type) : Type where 
 | empty : BinTree \alpha | node : \alpha \rightarrow BinTree \alpha \rightarrow BinTree \alpha
```

With binary trees, structural induction produces two induction hypotheses, one for each subtree of an inner node. To prove a goal  $t:BinTree\ \alpha\ \vdash\ P[t]$  by structural induction on t, we need to show the subgoals

```
\vdash P[BinTree.empty]
\mathbf{a}:\alpha,\mathbf{lr}:BinTree\alpha,\mathbf{ih\_l}:P[l],\mathbf{ih\_r}:P[r]\vdash P[BinTree.node\ a\ l\ r]
```

The tree counterpart to list reversal is the mirror operation:

```
def mirror {\alpha : Type} : BinTree \alpha \to BinTree \alpha | BinTree.empty => BinTree.empty | BinTree.node a l r => BinTree.node a (mirror r) (mirror l)
```

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Mirroring can be defined directly, without appealing to some append operation. As a result, reasoning about mirror is simpler than reasoning about reverse, as we can see below:

A more detailed informal proof would be as follows:

The proof is by structural induction on t.

CASE BinTree.empty: We must show that mirror (mirror BinTree. empty) = BinTree.empty. This follows directly from the definition of mirror.

CASE BinTree.node a l r: The induction hypotheses are

```
(ih_l) mirror (mirror l) = l (ih_r) mirror (mirror r) = r
We must show mirror (mirror (BinTree.node a l r)) = BinTree.node
a l r. We have
```

To achieve the same level of detail in the Lean proof, we could use a calculational block (Section 4.4) instead of simp:

The following theorem will be useful in Chapter 6:

```
theorem mirror_Eq_empty_Iff \{\alpha : Type\} : \forall t : BinTree \alpha, mirror t = BinTree.empty \leftrightarrow t = BinTree.empty | BinTree.empty | > by simp [mirror] | BinTree.node _ _ = > by simp [mirror]
```

### 5.9 Cases Tactic

#### cases

The cases tactic performs a case distinction on the specified term. This gives rise to as many subgoals as there are constructors in the definition of the term's type. The tactic is similar to induction except that it does not produce induction hypotheses and it automatically excludes impossible cases. The optional names  $names_1, ..., names_n$  are used for any emerging variables or hypotheses.

```
cases hypothesis-of-the-form-l-equals-r
```

The cases tactic can also be used on a hypothesis h of the form l = r. It matches r against l and replaces all occurrences of the variables occurring in r with the corresponding terms in l everywhere in the goal. The remaining hypothesis l = l can be removed using clear h if desired.

```
cases Classical.em (proposition) with
| inl name-if-true => tactics-if-true
| inr name-if-false => tactics-if-false
```

The cases tactic can also be used to perform a case distinction on a proposition. Two cases emerge: one in which the proposition is true and one in which it is false. The optional names <code>name-if-true</code> and <code>name-if-false</code> are used for the hypothesis in the true and false cases, respectively.

# 5.10 Dependent Inductive Types

The inductive types List  $\alpha$  and BinTree  $\alpha$  fall within the simply typed fragment of Lean. Inductive types may also depend on (non-type) terms. A typical example is the type of lists of length n, or *vectors*:

```
inductive Vec (\alpha : Type) : \mathbb{N} \to Type where | nil | : Vec \alpha 0 | cons (a : \alpha) {n : \mathbb{N}} (v : Vec \alpha n) : Vec \alpha (n + 1)
```

Thus, the term Vec.cons 3 (Vec.cons 1 Vec.nil) has type Vec  $\mathbb N$  2. By encoding the vector length in the type, we can provide more precise information about the result of functions. A function such as Vec.reverse, which reverses a vector, would map a value Vec  $\alpha$  n to another value of the same type, with the same n. And

Vec.zip could require its two arguments to have the same length. Fixed-length vectors and matrices are also useful in mathematics.

Unfortunately, this more precise information comes at a cost. Dependent inductive types cause difficulties when the terms they depend on are provably equal but not syntactically equal up to computation (e.g., m + n vs. n + m). In this guide, we will generally avoid dependent inductive types. They are briefly covered in this section for completeness. To put it unambiguously: This is not exam material.

The definitions below introduce conversions between lists and vectors:

```
def listOfVec {\alpha : Type} : \forall{n : \mathbb{N}}, Vec \alpha n \rightarrow List \alpha | _, Vec.nil => [] | _, Vec.cons a v => a :: listOfVec v def vecOfList {\alpha : Type} : \forallxs : List \alpha, Vec \alpha (List.length xs) | [] => Vec.nil | x :: xs => Vec.cons x (vecOfList xs)
```

The listOfVec conversion takes a type  $\alpha$ , a length n, and a vector of length n over  $\alpha$  as arguments and returns a list over  $\alpha$ . Although we do not care about the length n, it still needs to be an argument because it appears in the type of the third argument. We make the first two arguments,  $\alpha$  and n, implicit since they can be inferred from the type of the third argument.

The vecOfList conversion takes a type  $\alpha$  and a list over  $\alpha$  as arguments and returns a vector of the same length as the list. Lean's type checker is strong enough to determine that the two right-hand sides have the desired type.

By the PAT principle, proofs are analogous to function definitions. Let us verify that converting a vector to a list preserves its length:

To prove a goal  $v : Vec \alpha n \vdash P[v]$  by structural induction on v, we might naively think that it suffices to show the following two subgoals:

```
\vdash P[Vec.nil]
m: \mathbb{N}, a: \alpha, u: Vec \alpha m, ih: P[u] \vdash P[Vec.cons a u]
```

This is naive because the subgoals are not even type-correct: The hole in P[] has type Vec  $\alpha$  n (the type of its original dweller, v), so we cannot simply plug Vec.nil, u, or Vec.cons a u—which have types Vec  $\alpha$  0, Vec  $\alpha$  m, and Vec  $\alpha$  (m + 1)—into that hole. We must massage P each time, replacing n with 0, m, or m + 1. Using the notation P<sub>t</sub>[] for the variant of P[] where all occurrences of n are replaced by term t, we get

```
\vdash P_{0}[\text{Vec.nil}] m:\mathbb{N}, a:\alpha, u:\text{Vec}\ \alpha\, m, \text{ ih}:P_{m}[u]\vdash P_{m+1}[\text{Vec.cons}\ a\ u]
```

Proofs by case distinction using the cases tactic are similar, but without the induction hypothesis. Often, the length n will be not a variable but a complex

term. Then the replacement of n in P[] might not be intuitively meaningful. With cases, the corresponding subgoal is silently eliminated. Thus, a case distinction on a value of type  $Vec \ \alpha \ 0$  will yield only one subgoal, of the form  $\vdash P[Vec.nil]$ , since 0 could never be equal to a term of the form m + 1.

Dependently typed pattern matching is subtle, because the type of the value we match on may change according to the constructor. Given  $v: \text{Vec } \alpha$  n, we might be tempted to write

```
match v with
| Vec.nil => ...
| Vec.cons a u => ...
```

but this is just as naive as our first induction proof attempt above. Since the term n in the type  $Vec\ \alpha$  n may change depending on the constructor, we must pattern-match on n as well:

Showing the implicit arguments, we have

```
match n, v with | \circ, \quad \text{@Vec.nil } \alpha \quad => \dots | \text{m} + 1, \quad \text{@Vec.cons } \alpha \text{ a m u} => \dots
```

Often, it is sufficient to put placeholders in the first column:

It may seem paradoxical to pattern-match on n only to ignore the result, but without it Lean cannot infer the second implicit argument of Vec.cons. In this respect, cases is more user-friendly than match.

# **5.11 Summary of Newly Introduced Lean Constructs**

#### **Declarations**

class declares a structure type as a type class instance declares a structure value as a type class instance structure introduces a structure type and its selectors

#### **Expressions**

```
if ... then ... else performs a case distinction on a decidable proposition match ... with performs pattern matching
```

# Tactic

cases performs a case distinction

# **Chapter 6**

# **Inductive Predicates**

Inductive predicates, or inductively defined propositions, are a convenient way to specify functions of type  $\cdots \to \mathsf{Prop}$ . They are reminiscent of formal systems (Section 1.3) and of Prolog-style logic programming. But Lean is much more expressive than Prolog, so we need to do some work to establish theorems instead of just running the Prolog interpreter. A possible view of Lean:

Lean = functional programming + logic programming + more logic

## **6.1 Introductory Examples**

Unless you have been exposed to Prolog or logic programming, you will probably wonder what inductive predicates are and why they are useful. We start by reviewing three examples that demonstrate the variety of uses: even numbers, tennis games, and the reflexive transitive closure.

#### 6.1.1 Even Numbers

Mathematicians often define particular sets as the smallest set that meets some criteria. Consider this definition:

The set E of even natural numbers is defined as the smallest set S closed under the following rules:

- (1)  $0 \in S$ :
- (2) for every  $k \in \mathbb{N}$ , if  $k \in S$ , then  $k + 2 \in S$ .

Such a set exists by the Knaster-Tarski theorem.

(The last sentence is often left implicit.) It is easy to convince ourselves that *E* contains all the even numbers and only those. Let us put on our mathematician's hat and prove that 4 is even:

```
By rule (1), we have o \in E.
Hence, by rule (2) (with k := o), we have e \in E.
Thus, by rule (2) (with e := e), we have e \in E, as desired.
```

By contrast, computer scientists might use a formal system consisting of two derivation rules to specify the same set:

$$\frac{k \in E}{O \in E} \mathsf{ZERO} \qquad \frac{k \in E}{k + 2 \in E} \mathsf{ADDTWO}_{k}$$

A proof is then a derivation tree:

$$\frac{O \in E}{\frac{2 \in E}{4 \in E}} ADDTWO_{0}$$

$$\frac{ADDTWO_{2}}{4 \in E}$$

The proof is forward if we read it downwards and backward if we read it upwards. Inductive predicates are the logician's way to achieve the same result. In Lean, instead of a set, we would define a characteristic predicate inductively:

```
inductive Even : \mathbb{N} \to \operatorname{Prop} where | \operatorname{zero} : \operatorname{Even} \circ | add two : \forall k : \mathbb{N}, Even k \to \operatorname{Even} (k + 2)
```

This should look familiar. We have used the same syntax, except with Type instead of Prop, to define inductive types. Inductive types and inductive predicates are provided by the same mechanism in Lean, in accordance with the PAT principle.

The command above defines a unary predicate Even as well as two introduction rules Even.zero and Even.add\_two that can be used to prove goals of the form  $\vdash$  Even .... Recall that an introduction rule for a symbol (e.g., Even) is a theorem whose conclusion contains that symbol (Section 4.3). By the PAT principle, Even n can be viewed as a dependent inductive type like Vec  $\alpha$  n (Section 5.10), and Even.zero and Even.add\_two as constructors like Vec.nil and Vec.cons.

If we translate the Lean definition back into English, we get something similar to the Knaster–Tarski-style definition above:

The following clauses define the even numbers.

- (1) o is even;
- (2) any number of the form k + 2 is even, assuming k is even.

Any other number is not even.

It is worth nothing that inductively defined symbols have no definition. Thus, we cannot unfold their definition using simp. The only reasoning principles available are introduction and elimination.

As a warm-up exercise, here is a proof of Even 4:

```
theorem Even_4 :
   Even 4 :=
   have Even_0 : Even 0 :=
     Even.zero
   have Even_2 : Even 2 :=
     Even.add_two _ Even_0
   show Even 4 from
   Even.add_two _ Even_2
```

For example, the proof term Even.add\_two \_ Even\_o has type Even (o + 2), which is syntactically equal to Even 2 up to computation and hence equal from the type checker's point of view. The underscore stands for o.

Thanks to the "no junk" guarantee of inductive definitions, Even.zero and Even.add\_two are the only two ways to construct proofs of  $\vdash$  Even... By inspecting the conclusions Even o and Even (k + 2), we see that there is no risk of ever proving that 1 is even.

Another way to view the inductive definition above is as follows: The first line introduces a predicate, whereas the second and third lines introduce axioms we want the predicate to satisfy. Accordingly, we could have written

```
opaque Even : \mathbb{N} \to \mathsf{Prop} axiom Even.zero : Even o axiom Even.add_two : \forall k : \mathbb{N}, Even k \to \mathsf{Even} (k + 2)
```

replacing inductive by opaque and | by axiom. But this axiomatic version does not give us any information about when Even is false. We cannot use it to prove  $\neg$  Even 1 or  $\neg$  Even 17. For all we know, Even could be true for all natural numbers. In contrast, the inductive definition guarantees that we obtain the least (i.e., the most false) predicate that satisfies the introduction rules Even.zero and Even.add\_two, and provides elimination and induction principles that allow us to prove  $\neg$  Even 1,  $\neg$  Even 17, or  $\neg$  Even (2 \* n + 1).

Why should we bother with inductive predicates when we can define recursive functions? Indeed, the following definition is perfectly legitimate:

Each style has its strengths and weaknesses. The recursive version forces us to specify a false case (the second equation), and it forces us to worry about termination. On the other hand, because it is equational and computational, it works well with rfl, simp, #eval, and #reduce. The inductive version is arguably more abstract and elegant. Each introduction rule is stated independently. We can add or remove rules without having to think about termination or executability.

Yet another way to define Even is as a nonrecursive definition, using the modulo operator (%):

```
def evenNonrec (k : N) : Prop :=
  k % 2 = 0
```

Mathematicians would likely prefer this version. But the inductive version is a convenient "Hello, World!" example that resembles many realistic inductive definitions. It might be a toy, but it is a useful toy.

#### 6.1.2 Tennis Games

Transition systems consists of transition rules that connect a "before" and an "after" state. As a specimen of a transition system, we consider the possible transitions in a game of tennis, starting from 0–0 ("Love all"). Tennis games are also a toy, but in Chapter 9, we will define the semantics of an imperative programming language as a transition system in a similar style.

The scoring rules for tennis from the International Tennis Federation's *Rules* of *Tennis* are reproduced below.

A standard game is scored as follows with the server's score being called first:

```
No point - "Love"
First point - "15"
Second point - "30"
Third point - "40"
Fourth point - "Game"
```

except that if each player/team has won three points, the score is "Deuce." After "Deuce," the score is "Advantage" for the player/team who wins the next point. If that same player/team also wins the next point, that player/team wins the "Game"; if the opposing player/team wins the next point, the score is again "Deuce." A player/team needs to win two consecutive points immediately after "Deuce" to win the "Game."

We first define an inductive type to represent scores:

A score such as 30–15 is represented as Score.vs 30 15, which we will also write in infix notation as 30–15. We ignore some of the most frivolous aspects of the scoring rules, writing 0 for "Love" and 40-40 for "Deuce." If we wanted to, we could introduce aliases such as def love:  $\mathbb{N}$  := Score.vs 0 0.

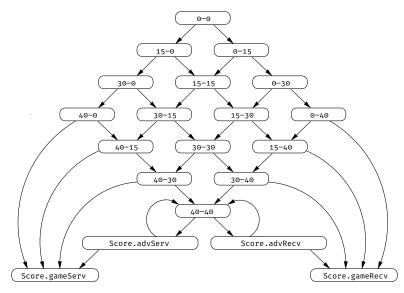
The next stage is to introduce a binary predicate Step that determines whether a transition is possible:

Let  $s \rightsquigarrow t$  abbreviate Step s t. A game is a chain  $s_0 \rightsquigarrow s_1 \rightsquigarrow s_2 \rightsquigarrow \cdots \rightsquigarrow s_n$  where  $s_0 = o-o$  and no transition is possible from  $s_n$ . The predicate allows nonsensical transitions such as  $15-99 \rightsquigarrow 30-99$ , but since the score 15-99 cannot be reached from o-o, these transitions are harmless.

Equipped with a formal definition, we can ask, and formally answer, questions such as: Do games have a maximal length? How many different final scores are possible? Is the score 15–99 reachable from "Love all"? And can the score ever return to 0–0? Let us use Lean to disprove the last claim:

```
theorem no_Step_to_o_o (s : Score) :
  ¬ s ~> o-o :=
  by
    intro h
    cases h
```

The diagram below summarizes which scores are reachable from which scores:



### 6.1.3 Reflexive Transitive Closure

As a particularly useful application of inductive predicates, consider the reflexive transitive closure  $r^*$  of a binary relation r. Informally,  $r^*$  is the relation that represents zero or more steps of r. For example, if r corresponds to taking one transition in a transition system (e.g., an automaton), then  $r^*$  corresponds to taking any number of steps, including zero. For a more concrete example, consider the following:

$$r = \{(1,2),(2,4),(4,8)\}$$

$$r^* = \{(n,n) \mid n \in \mathbb{N}\} \cup \{(1,2),(1,4),(1,8),(2,4),(2,8),(4,8)\}$$

The star (\*) operator is often defined rigorously as a formal system:

$$\frac{(a,b) \in r}{(a,b) \in r^*} \mathsf{BASE} \qquad \frac{(a,b) \in r^*}{(a,a) \in r^*} \mathsf{REFL} \qquad \frac{(a,b) \in r^*}{(a,c) \in r^*} \mathsf{TRANS}$$

These rules define  $r^*$  as the smallest relation that contains r (by BASE) and that is reflexive (by REFL) and transitive (by TRANS). If we wanted to define the transitive

closure  $r^*$  instead, we would simply omit the REFL rule. If we wanted the reflexive symmetric closure, we would replace the TRANS rule by a SYMM rule. With a formal system, we simply declare the properties we want to be true, without giving a thought to termination or executability.

It is straightforward to translate the above derivation rules into the introduction rules of an inductive predicate:

```
inductive Star \{\alpha: \mathsf{Type}\}\ (\mathsf{R}: \alpha \to \alpha \to \mathsf{Prop}): \alpha \to \alpha \to \mathsf{Prop} where |\ \mathsf{base}\ (\mathsf{a}\ \mathsf{b}: \alpha) \qquad : \ \mathsf{R}\ \mathsf{a}\ \mathsf{b} \to \mathsf{Star}\ \mathsf{R}\ \mathsf{a}\ \mathsf{b} \\ |\ \mathsf{refl}\ (\mathsf{a}: \alpha) \qquad : \ \mathsf{Star}\ \mathsf{R}\ \mathsf{a}\ \mathsf{a} \\ |\ \mathsf{trans}\ (\mathsf{a}\ \mathsf{b}\ \mathsf{c}: \alpha): \ \mathsf{Star}\ \mathsf{R}\ \mathsf{a}\ \mathsf{b} \to \mathsf{Star}\ \mathsf{R}\ \mathsf{b}\ \mathsf{c} \to \mathsf{Star}\ \mathsf{R}\ \mathsf{a}\ \mathsf{c}
```

We represent relations as binary predicates rather than sets of pairs, writing R a b for  $(a,b) \in R$ . Relations and predicates are interchangeable, but predicates are often more convenient to use in a proof assistant. The reflexive transitive closure of R is written Star R. Notice that a, b, and c are declared as parameters of the introduction rules, on the left of the colons. We could also have written

All these forms are logically and operationally equivalent.

The general format of inductive predicates is as follows:

```
inductive predicate-name (params_1: type_1) ... (params_k: type_k): type_{k+1} \rightarrow \cdots \rightarrow type_{k+p} \rightarrow Prop \ where
| rule-name_1 (params_{11}: type_{11}) ... (params_{1m_1}: type_{1m_1}): proposition_1
:
| rule-name_n (params_{n1}: type_{n1}) ... (params_{nm_n}: type_{nm_n}): proposition_n
```

where the conclusion of each  $proposition_j$  must be an application of the defined predicate predicate-name to some arguments. These arguments may be arbitrary terms; they do not need to be constructor patterns. We can also use curly braces  $\{\ \}$  instead of parentheses ( ) if we want to make the arguments corresponding to the parameters implicit.

The above definition of Star is truly elegant. If you still doubt this, try implementing it as a recursive function:

```
def starRec {\alpha : Type} (R : \alpha \to \alpha \to \text{Bool}) : \alpha \to \alpha \to \text{Bool} :=
```

To summarize, each introduction rule of an inductive predicate P consists of the following components, from left to right:

- a name;
- variables that may appear in the rule;
- zero or more conditions that must be fulfilled, which may apply P recursively;
- an application of P to some arguments, forming a pattern.

Accordingly, for rule Star.base, the pattern is Star R a b, the condition is R a b, and the variables are a and b.

#### 6.1.4 A Nonexample

Not all inductive definitions admit a least solution. The simplest nonexample is

```
-- fails
inductive Illegal : Prop where
| intro : ¬ Illegal → Illegal
```

If Lean accepted this definition, we could use it to prove the equivalence Illegal  $\leftrightarrow \neg$  Illegal, from which we could easily derive False. Fortunately, Lean rejects the definition:

```
arg #1 of 'Illegal.intro' has a non positive occurrence of
the datatypes being declared
```

The nonpositive occurrence it complains about is the occurrence of Illegal under a negation. Mathematicians would reject the definition on the ground that the monotonicity condition of the Knaster-Tarski theorem is not satisfied.

# 6.2 Logical Symbols

Although Even is the first openly inductive predicate in this guide, the earlier chapters already presented other inductive predicates clandestinely. The very first of these was equality (=), introduced in Chapter 2, followed by the logical symbols  $\land$ ,  $\lor$ ,  $\leftrightarrow$ ,  $\exists$ , True, and False. Their definitions are worth studying closely:

```
inductive And (a b : Prop) : Prop where | intro : a \rightarrow b \rightarrow And a b | inductive Or (a b : Prop) : Prop where | inl : a \rightarrow Or a b | inr : b \rightarrow Or a b | intro : (a b : Prop) : Prop where | intro : (a b : Prop) : Prop where | intro : (a b : Prop) : Prop where | intro : \foralla : \alpha, Pa \rightarrow Exists P | inductive True : Prop where | intro : True | inductive False : Prop where | inductive Eq {\alpha : Type} : \alpha \rightarrow \alpha \rightarrow Prop where | inductive Eq {\alpha : Type} : \alpha \rightarrow \alpha \rightarrow Prop where | refl : \foralla : \alpha, Eq a a
```

(Strictly speaking, in Lean, some of the above definitions are actually structures and not inductive predicates, but the difference is superficial. As we saw in Section 5.5, structures are essentially single-constructor inductive predicates with some syntactic sugar.)

The traditional notations  $\exists x: \alpha$ , P and x = y are syntactic sugar for Exists (fun  $x: \alpha \mapsto P$ ) and Eq x y. Notice how fun acts as an all-purpose binder. Notice also that there is no constructor for False. There are no proofs of False, just like there is no proof of Even 1. With inductive predicates, we state only the rules we want to be true.

The symbol  $\forall$ , including its special case  $\rightarrow$ , is built directly into the logic and is not defined as an inductive predicate. Nor does it have explicit introduction or elimination rules. The introduction principle is the anonymous function fun  $x \mapsto$  \_, and the elimination principle is function application \_ u.

As for any inductive predicates, only the introduction rules are specified. The elimination rules presented in Sections 3.3 and 3.4 must be derived manually.

### 6.3 Rule Induction

In the same way that we can perform an induction on a term of inductive type, we can perform an induction on a proof of an inductive predicate. For example, given the goal  $h : Even n \vdash P n$ , we can invoke induction h and get two subgoals, for Even.zero and Even.add\_two. This is called *induction on the structure of the derivation of h* or simply *rule induction*, because the induction is on the predicate's introduction rules (i.e., the constructors of the proof term).

There are two ways to look at rule induction: the "least-predicate-such-that view" and the "PAT view." To understand the least-predicate-such-that view, recall that an inductive definition introduces a symbol as the least (i.e., the most false) predicate satisfying the introduction rules. Accordingly, Even is the least predicate Q such that the properties Q o and  $\forall k$ , Q k  $\rightarrow$  Q (k + 2) hold. Therefore, if we can show that P o and  $\forall k$ , P k  $\rightarrow$  P (k + 2) hold for some predicate P, then P is either Even itself or greater than (i.e., more true than) Even. As a result, Even n implies P n, which is exactly what we need to prove the goal h: Even n  $\vdash$  P n.

The least-predicate-such-that view gives a nice intuitive account of rule induction that can be used in informal arguments, such as the following proof that Even n implies n % 2 = 0 for all n:

The proof is by rule induction on the hypothesis Even n.

CASE Even. zero: We must show 0 % 2 = 0. This follows by computation.

CASE Even.add\_two k: The induction hypothesis is k % 2 = 0. We must show (k + 2) % 2 = 0. This follows by basic arithmetic reasoning.

The Lean proof has the same structure:

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The PAT principle gives us another fruitful way to look at rule induction. The key idea is that rule induction on h in a goal such as h: Even  $n \vdash P[h]$  is perfectly analogous to structural induction on a value of a dependent inductive type such as Vec  $\alpha$  n (Section 5.10). Writing  $P_u[\ ]$  for the variant of  $P[\ ]$  where n is replaced by some term u, we get the subgoals

```
\vdash P_0[\text{Even.zero}: \text{Even o}]
k: \mathbb{N}, hk: Even k, ih: P_k[\text{hk}] \vdash P_{k+2}[\text{Even.add\_two k hk}: \text{Even (k + 2)}]
```

These are effectively the subgoals produced by the induction tactic.

Regardless of the inductive predicate Q, the procedure to compute the subgoals is always the same:

- Replace h in P[h] with each possible introduction rule applied to fresh variables (e.g., Even.add\_two k hk), instantiating n in P[] to make P[] type-correct. This yields as many subgoals as there are introduction rules.
- 2. Add these new variables (e.g., k, hk) to the local context.
- 3. Add induction hypotheses for all new hypotheses that assert Q . . . .

Notice the presence of hk: Even k among the hypotheses above. It was absent in the least-predicate-such-that view and is not essential because it can always be recovered by strengthening P to be of the form Even  $n \wedge \cdots$ .

In nearly all practical cases, h will not occur in P[h]. We can then simply write

```
\vdash P_0   k : \mathbb{N}, hk : Even k, ih : P_k \vdash P_{k+2}
```

In rare cases, h will occur in P[h]. Proofs may appear as subterms in arbitrary terms, as we saw when we tried to extract the head of a list in Section 5.7.

The reflexive transitive closure Star R is similar to Even. Given a goal  $h: Star R \times y \vdash P$ , rule induction on h produces the following subgoals, where  $P_{t,u}$  denotes the variant of P where x and y are replaced by t and u, respectively:

```
\begin{array}{c} \textbf{a} \ \textbf{b} : \alpha, \ \textbf{hab} : \textbf{R} \ \textbf{a} \ \textbf{b} \ \vdash \ \textbf{P}_{\textbf{a},\textbf{b}} \\ & \textbf{a} : \alpha \vdash \ \textbf{P}_{\textbf{a},\textbf{a}} \\ \\ \textbf{a} \ \textbf{b} \ \textbf{c} : \alpha, \ \textbf{hab} : \textbf{Star} \ \textbf{R} \ \textbf{a} \ \textbf{b}, \ \textbf{hbc} : \textbf{Star} \ \textbf{R} \ \textbf{b} \ \textbf{c}, \ \textbf{ihbc} : \textbf{P}_{\textbf{a},\textbf{b}}, \ \textbf{ihbc} : \textbf{P}_{\textbf{b},\textbf{c}} \vdash \ \textbf{P}_{\textbf{a},\textbf{c}} \end{array}
```

This is where the "assistant" aspect of "proof assistant" comes into play. One of the key properties of Star is idempotence—applying Star to Star R has no effect. This can be proved as follows in Lean, using rule induction for the  $\rightarrow$  direction of the equivalence:

```
apply Star.trans a b
  { exact ihab }
  { exact ihbc } }
{ intro h
  apply Star.base
  exact h }
```

We are careful to give intuitive names to the emerging variables. It is easy to get lost in goals containing long, automatically generated names. The cleanup tactics introduced in Section 3.8 can also be of help when we face large goals.

We can state the idempotence property more conventionally in terms of equality instead of as an equivalence:

```
@[simp] theorem Star_Star_Eq_Star {α : Type}
    (R : α → α → Prop) :
    Star (Star R) = Star R :=
    by
        apply funext
    intro a
        apply funext
    intro b
        apply propext
        apply Star_Star_Iff_Star
```

The proof uses two theorems that are available because Lean's logic is classical:

```
funext: (\forall x, ?f x = ?g x) \rightarrow ?f = ?g
propext: (?a \leftrightarrow ?b) \rightarrow ?a = ?b
```

Functional extensionality (funext) states that if two functions yield equal results for all inputs, then the two functions must be equal. Propositional extensionality (propext) states that equivalence of propositions coincides with equality. In these phrases, extensionality means something like "what you see is what you get." Although these properties may seem obvious, there are some proof assistants built on weaker, intuitionistic logics in which the properties do not generally hold.

We register the theorem Star\_Star\_Eq\_Star as a simp rule, because viewed as a left-to-right rewrite rule, it genuinely replaces a complex term by a simpler term. It is hard to imagine a situation where we would not want simp to rewrite Star (Star...) to Star....

For rule induction, we use the induction tactic. For subtle logical reasons that will become clearer in Chapter 12, rule induction by pattern matching and recursion is not allowed.

In Section 5.4, we saw a diagram depicting the interpretation of Bool and Propside by side. The diagram suggested the existence of an infinite number of propositions, but we now know that there are exactly two propositions: False and True. Here is a revised diagram:



We see that there are only two propositions, one which has no proofs and one which has a number of proofs. Three proofs are shown in the diagram. We will refine this picture again in Chapter 12.

### 6.4 Linear Arithmetic Tactic

#### linarith

The linarith tactic can be used to prove goals involving linear arithmetic equalities (=), inequalities (<, >,  $\leq$ , and  $\geq$ ), and disequalities ( $\neq$ ). Linear means that multiplication and division do not occur, or if they do then one of the operands must be a numeric constant. For example, 2 \* x < y is a linear constraint (which can be rewritten to x + x < y), whereas x \* y < y is nonlinear.

### 6.5 Elimination

Given an inductive predicate Q, its introduction rules typically are of the form  $\forall \ldots, \cdots \to Q \ldots$  and can be used to prove goals of the form  $\vdash Q \ldots$  Elimination works the other way around: It extracts information from a theorem or hypothesis  $h:Q\ldots$  Elimination takes many forms: the cases and induction tactics, pattern matching, and elimination rules (e.g., And.left).

Invoked on h: Q..., the cases h tactic performs roughly the same rule induction as induction h but without producing any induction hypotheses. We encountered two idioms in Chapter 5 that we can finally analyze.

The first idiom is when h is of the form l = r—i.e., Eq l r (Section 6.2). Suppose the goal is  $h : l = r \vdash P[h]$ . The procedure presented in Section 6.3 produces the subgoal

$$a: \alpha \vdash P_{a,a}[Eq.refl a: a = a]$$

in which  $P_{t,u}[$  ] stands for the variant of P[ ] where l and r are replaced by t and u, respectively. (Strictly speaking, the useless hypothesis h:a=a would also appear in the subgoal.) In practice, P[h] would likely not depend on h. Moreover, cases reuses the name l instead of using a different name like a. Thus, we would get

$$l: \alpha \vdash P_{1.1}$$

In other words, all occurrences of r in the original goal have been replaced by 1. This corresponds to the behavior we observed in Section 5.7.

The second idiom is the tactic cases Classical.em Q, where Q is a proposition. The Classical.em Q part is a proof term for  $Q \lor \neg Q$ —i.e.,  $Q \lor Q \lor Q$  (Section 6.2). Then cases is applied to eliminate the  $\lor Q \lor Q$  connective. Suppose the goal is  $\vdash Q \lor Q$  Classical.em Q]. By the definition of the  $Q \lor Q \lor Q$ 

```
hQ : Q \vdash P[Or.inl hQ : Q \lor \neg Q]

hnQ : \neg Q \vdash P[Or.inr hnQ : Q \lor \neg Q]
```

There is no need to modify P because Or.inl hQ and Or.inr hnQ have the same type as Classical.em Q—namely,  $Q \lor \neg Q$ . In practice, the goal would usually not contain Classical.em Q and be simply  $\vdash P$ , and we would then have the subgoals

$$hQ:Q\vdash P$$
  $hnQ:\neg Q\vdash P$ 

Again, this is the behavior we observed in Section 5.7.

In structured proofs, we can use match expressions (Section 5.4) to achieve the same effect as cases. This works well for logical symbols. However, for predicates such as Even and Star, with arguments that evolve through the induction, we end up with dependently typed pattern matching, which is subtle (Section 5.10). It is generally easier to let cases figure out what the subgoals should look like than to pattern-match. We will review an example of both styles below.

It can be useful to expand a hypothesis of the form Q(c...), where c is a constructor or some other constant. We can state and prove an *inversion rule* to support such eliminative reasoning. A typical inversion rule has the form

```
\forall x_1 \dots x_n, Q (c x_1 \dots x_n) \rightarrow (\exists \dots, \dots \land \dots) \lor \dots \lor (\exists \dots, \dots \land \dots)
```

It can be useful to combine introduction and elimination into a single theorem, which can be used for rewriting both the hypotheses and the targets of goals. The format is the same except for the connective  $\leftrightarrow$  in the middle:

```
\forall x_1 \dots x_n, Q (c x_1 \dots x_n) \leftrightarrow (\exists \dots, \dots \land \dots) \lor \dots \lor (\exists \dots, \dots \land \dots)
```

An inversion rule for Even is given below:

```
theorem Even_Iff (n : N) :
  Even n \leftrightarrow n = 0 \lor (\exists m : \mathbb{N}, n = m + 2 \land \text{Even m}) :=
  by
    apply Iff.intro
    { intro hn
      cases hn with
      lzero
                       => simp
      | add two k hk =>
         apply Or.inr
         apply Exists.intro k
         simp [hk] }
    { intro hor
      cases hor with
      | inl heq => simp [heq, Even.zero]
      | inr hex =>
         cases hex with
         | intro k hand =>
           cases hand with
           | intro heq hk =>
             simp [heq, Even.add_two _ hk] }
```

As usual, the tactical proof is not particularly readable, but we see that introduction rules and the eliminative cases tactic play a major role, for both the logical symbols and the Even predicate. The simp tactic puts the final touches.

If you prefer structured proofs, here is a version of the proof with dependently typed pattern matching on hn: Even n:

```
theorem Even_Iff_struct (n : \mathbb{N}) : Even n \leftrightarrow n = 0 \vee (\existsm : \mathbb{N}, n = m + 2 \wedge Even m) := Iff.intro (assume hn : Even n
```

```
match n, hn with
 | _, Even.zero
                          =>
   show \odot = \odot \lor _ from
     by simp
 | _, Even.add_two k hk =>
   show \_ \lor (\exists m, k + 2 = m + 2 \land Even m) from
     Or.inr (Exists.intro k (by simp [*])))
(assume hor : n = 0 \lor (\exists m, n = m + 2 \land Even m)
 match hor with
 | Or.inl heg =>
   show Even n from
     by simp [heq, Even.zero]
 | Or.inr hex =>
   match hex with
   | Exists.intro m hand =>
     match hand with
     | And.intro heq hm =>
       show Even n from
          by simp [heq, Even.add_two _ hm])
```

## 6.6 Further Examples

Equipped with a better understanding of inductive predicates, we are now ready to review four further applications in turn.

#### 6.6.1 Sorted Lists

Our first example is a predicate that checks whether a list of natural numbers is sorted in increasing order:

```
inductive Sorted : List N → Prop where
  | nil : Sorted []
  | single (x : N) : Sorted [x]
  | two_or_more (x y : N) {zs : List N} (hle : x ≤ y)
        (hsorted : Sorted (y :: zs)) :
        Sorted (x :: y :: zs)
```

This definition captures the following mathematical intuition:

The set of sorted lists is defined as the smallest set closed under the following rules:

- (1) the list [] is sorted;
- (2) given a number x, the list [x] is sorted;
- (3) given two numbers x, y and a list zs, if x < y and y :: zs is sorted, then x :: y :: zs is sorted.

It is always a good idea to test our definitions by trying it on small examples. Is the list [3, 5] sorted? It would appear so:

```
theorem Sorted_3_5 :
   Sorted [3, 5] :=
   by
     apply Sorted.two_or_more
   { simp }
   { exact Sorted.single _ }
```

The example needs two of the introduction rules for sorted. This approach tends to work well for expressions consisting only of closed terms. A more compact proof follows, using proof terms:

```
theorem Sorted_3_5_raw :
  Sorted [3, 5] :=
  Sorted.two_or_more _ _ (by simp) (Sorted.single _)
```

The same idea can be used to prove that [7, 9, 9, 11] is sorted:

```
theorem sorted_7_9_9_11 :
   Sorted [7, 9, 9, 11] :=
   Sorted.two_or_more _ _ (by simp)
      (Sorted.two_or_more _ _ (by simp)
      (Sorted.two_or_more _ _ (by simp)
      (Sorted.single _)))
```

Conversely, we can show that some lists are not sorted. For this, we need elimination:

```
theorem Not_Sorted_17_13 :
    ¬ Sorted [17, 13] :=
    by
    intro h
    cases h with
    | two_or_more _ _ hlet hsorted => simp at hlet
```

From the hypothesis that the list [17, 13] is sorted, we extract the inequality 17  $\leq$  13. The cases tactic silently eliminates the nil and single cases, because they cannot match a two-element list. Then we invoke simp to exploit the impossible hypothesis 17  $\leq$  13.

#### 6.6.2 Palindromes

Palindromes are lists that read the same from left to right and from right to left. For example, [a, b, b, a] and [a, h, a] are palindromes. The following inductive predicate is True if and only if the list passed as argument is a palindrome:

```
inductive Palindrome \{\alpha: \mathsf{Type}\}: \mathsf{List}\ \alpha \to \mathsf{Prop}\ \mathsf{where} | nil : Palindrome [] | single (\mathsf{x}:\alpha): \mathsf{Palindrome}\ [\mathsf{x}] | sandwich (\mathsf{x}:\alpha)\ (\mathsf{xs}: \mathsf{List}\ \alpha)\ (\mathsf{hxs}: \mathsf{Palindrome}\ \mathsf{xs}): \mathsf{Palindrome}\ ([\mathsf{x}] ++ \ \mathsf{xs} \ ++ \ [\mathsf{x}])
```

The definition distinguishes three cases: (1) [] is a palindrome; (2) for any element x, the singleton list [x] is a palindrome; (3) for any element x and any palindrome [ $y_1, \ldots, y_n$ ], the list [ $x, y_1, \ldots, y_n, x$ ] is a palindrome.

Palindromes are another example where inductive predicates come into their own. The following naive recursive definition cannot work because [x] ++ xs ++ [x] is not a constructor pattern, and the variable x is repeated:

```
-- fails  \begin{array}{lll} \text{def palindromeRec } \{\alpha: \mathsf{Type}\} : \mathsf{List} \ \alpha \to \mathsf{Bool} \\ | \ [] & => \mathsf{true} \\ | \ [\_] & => \mathsf{true} \\ | \ ([\mathsf{x}] \ ++ \ \mathsf{xs} \ ++ \ [\mathsf{x}]) \ => \mathsf{palindromeRec} \ \mathsf{xs} \\ | & => \mathsf{false} \\ \end{array}
```

A correct recursive definition is possible but beyond the scope of this guide. Naturally, the reverse of a palindrome is a palindrome. It is a good exercise:

Informally:

The proof is by rule induction on the hypothesis hxs.

CASE Palindrome.nil: We must show Palindrome (reverse []). This follows from Palindrome.nil using reverse [] = [].

CASE Palindrome.single x: We must show Palindrome (reverse [x]). This follows from Palindrome.single using reverse [x] = [x].

CASE Palindrome.sandwich x xs hxs: We must show Palindrome (reverse ([x] ++ xs ++ [x])) under the hypothesis (hxs) Palindrome xs. The induction hypothesis is Palindrome (reverse xs). By simplification, it suffices to show Palindrome ([x] ++ reverse xs ++ [x]). By Palindrome.sandwich, it suffices to show Palindrome (reverse xs), which is exactly the induction hypothesis.  $\Box$ 

### 6.6.3 Full Binary Trees

Our third example is based on the type of binary trees introduced in Section 5.8:

```
inductive BinTree (\alpha : Type) : Type where 
| empty : BinTree \alpha | node : \alpha \to BinTree \alpha \to BinTree \alpha
```

A binary tree is *full* if all its nodes have either zero or two children. This can be encoded as an inductive predicate:

```
inductive IsFull \{\alpha: \mathsf{Type}\}: \mathsf{BinTree}\ \alpha \to \mathsf{Prop}\ \mathsf{where} | empty : IsFull BinTree.empty
```

```
| node (a : \alpha) (l r : BinTree \alpha)

(hl : IsFull l) (hr : IsFull r)

(hiff : l = BinTree.empty \leftrightarrow r = BinTree.empty) :

IsFull (BinTree.node a l r)
```

The first case states that the empty tree is a full tree. The second case states that a nonempty tree is a full tree if it has two child trees that are themselves full and that are both empty or both nonempty. The two cases neatly follow the structure of the inductive type, so it is natural to reuse the names empty and node.

The tree that consists of a node with empty trees as children is a full tree. Here is a simple proof:

```
theorem IsFull_singleton {α : Type} (a : α) :
   IsFull (BinTree.node a BinTree.empty BinTree.empty) :=
   by
   apply IsFull.node
   { exact IsFull.empty }
   { exact IsFull.empty }
   { rfl }
```

A somewhat more interesting property of full trees is that fullness is preserved by the mirror operation. Our first proof is by rule induction on ht: IsFull t:

Since IsFull's definition follows BinTree's definition, it is also reasonable to perform structural induction on the tree t:

```
apply IsFull.node
{ exact ih_r hr }
{ apply ih_l hl }
{ simp [mirror_Eq_empty_Iff, *] } }
```

The key here is the case distinction on the hypothesis ht: IsFull(BinTree.node a l r). The cases tactic notices that the IsFull.empty introduction rule cannot have been used to derive ht, so it only produces one case, corresponding to IsFull.node. As usual, the tactical proof will make more sense if you inspect it in Visual Studio Code, moving the cursor around.

### 6.6.4 First-Order Terms

Our last example is based on an inductive type of first-order terms:

```
inductive Term (\alpha \ \beta : \mathsf{Type}) : \mathsf{Type} where | \mathsf{var} : \beta \to \mathsf{Term} \ \alpha \ \beta | \mathsf{fn} : \alpha \to \mathsf{List} \ (\mathsf{Term} \ \alpha \ \beta) \to \mathsf{Term} \ \alpha \ \beta
```

A first-order term is either a variable x or a function symbol f applied to a list of arguments:  $f(t_1, \ldots, t_n)$ , where the mathematical variables  $t_1, \ldots, t_n$  stand for the arguments, which are themselves terms. Thus,  $\sin(\max(x, y))$  is a first-order term. The parameters  $\alpha$  and  $\beta$  are the types of function symbols and variables, respectively.

Not all terms are legal. For example, the term  $\min(\cos(a), \cos(a, b))$  is considered ill-formed, because the function cos is invoked with inconsistent number of arguments (1 versus 2). Along with  $\alpha$  and  $\beta$ , we also consider the arity, represented by a function  $\operatorname{arity}: \alpha \to \mathbb{N}$  indicating how many arguments each function symbol takes. For example, a binary symbol has arity 2.

The WellFormed predicate then checks whether the given term only contains function symbol applications with the specified number of arguments:

```
inductive WellFormed \{\alpha \ \beta : \text{Type}\}\ (\text{arity} : \alpha \to \mathbb{N}) : Term \alpha \ \beta \to \text{Prop where} | var (x : \beta) : \text{WellFormed arity (Term.var } x) | fn (f : \alpha) (ts : \text{List (Term } \alpha \ \beta)) (\text{hargs} : \forall t \in \text{ts, WellFormed arity } t) (\text{hlen} : \text{length } ts = \text{arity } f) : WellFormed arity (\text{Term.fn } f \ ts)
```

The fn case checks that the arguments ts are recursively well formed and that the length of ts equals the specified arity for the function symbol f in question.

Another interesting property of first-order terms is whether they contain variables. This can be checked easily using an inductive predicate:

```
inductive VariableFree \{\alpha \ \beta : \ \mathsf{Type}\} : \ \mathsf{Term} \ \alpha \ \beta \to \mathsf{Prop} \ \mathsf{where} \ | \ \mathsf{fn} \ (\mathsf{f} : \alpha) \ (\mathsf{ts} : \ \mathsf{List} \ (\mathsf{Term} \ \alpha \ \beta)) \ (\mathsf{hargs} : \ \forall \mathsf{t} \in \ \mathsf{ts}, \ \mathsf{VariableFree} \ \mathsf{t}) : \ \mathsf{VariableFree} \ (\mathsf{Term.fn} \ \mathsf{f} \ \mathsf{ts})
```

There is no introduction rule corresponding to Term.var because variables are never variable-free.

### 6.7 Induction Pitfalls

Some care is needed when invoking induction on inductive predicates. The arguments of inductive predicates often evolve through the induction. Such details are often glossed over in informal proofs, but proof assistants require us to be precise.

Recall the definition of even numbers:

```
inductive Even : \mathbb{N} \to \operatorname{Prop} where | \operatorname{zero} : \operatorname{Even} \circ | add two : \forall k : \mathbb{N}, Even k \to \operatorname{Even} (k + 2)
```

If the goal has the form  $h : Even n \vdash P n$ , applying induction on h will produce the following subgoals:

```
\vdash P \circ k : \mathbb{N}, hk : Pk \vdash P(k+2)
```

This works as desired.

The problem is when Even's argument is not a variable. Applying induction on the hypothesis hev in the goal hev: Even  $(2 * n + 1) \vdash False$  fails with an error:

```
index in target's type is not a variable (consider using
the 'cases' tactic instead)
```

To solve this issue, we need to replace 2 \* n + 1 by a variable m and add an equation m = 2 \* n + 1 as a hypothesis:

```
m: 2 * n + 1, hev: Even m ⊢ False
```

This goal is logically equivalent, but now induction produces two subgoals:

Sadly, the second subgoal is not provable. The problem is that we want to instantiate the n in the induction hypothesis ih with n-1 but n is not instantiable.

The solution? We need to universally quantify over n to be able to instantiate it in the induction hypothesis. This is achieved by specifying generalizing n after induction h. Now we obtain the subgoals

Now the variable n in the conclusion is disconnected from the variable n in the induction hypothesis, and we can instantiate the hypothesis's n with the conclusion's n-1.

Putting all of this together, we obtain

```
theorem Not_Even_two_mul_add_one (m n : N)
    (hm : m = 2 * n + 1) :
    The Even m :=
    by
    intro h
```

We use the theorem  $Nat.succ\_eq\_add\_one$  to rewrite terms of the form Nat.succ n to n + 1 and the linarith tactic to perform simple arithmetic reasoning. The theorem is very useful when we face a mixture of Nat.succ and addition.

# **6.8 Summary of Newly Introduced Lean Constructs**

#### **Theorems**

funext functional extensionality propext propositional extensionality

#### **Tactic**

linarith invokes a procedure for linear arithmetic

## **Chapter 7**

# **Effectful Programming**

Pure functional programming can sometimes feel overly restrictive. *Effectful functional programming* provides idioms that alleviate some of these restrictions, giving us the impression of programming with side effects, exceptions, nondeterminism, and other effects.

The underlying abstraction is called *monad*. Monads generalize programs with effects. They are popular in Haskell to write imperative programs. In Lean, they are used to express imperative programs and to reason about them. They are even useful for programming Lean itself, as we will see in Chapter 8.

These notes are inspired by Chapter 7 of *Programming in Lean* [2]. We also refer to Chapter 14 of *Real World Haskell* [24] for a general introduction to effectful functional programming.

## 7.1 Introductory Example

Consider the following programming task:

Implement a function sum257 ns that sums up the second, fifth, and seventh items of a list ns of natural numbers. Use Option  $\mathbb N$  for the result so that you can return Option.none if the list has too few items.

A straightforward solution would be as follows:

(Confusingly, nth counts elements from 0.) The code is quite inelegant, because of all the pattern matching on Option. Although the programming task is contrived, we can all recall writing code with nested error handling and ever increasing indentation levels.

We can do better, by concentrating the ugliness in one function:

```
def connect \{\alpha: \mathsf{Type}\} \{\beta: \mathsf{Type}\}:
Option \alpha \to (\alpha \to \mathsf{Option}\ \beta) \to \mathsf{Option}\ \beta
| Option.none, _ => Option.none
| Option.some a, f => f a
```

The connect function works on an Option. If the value is Option.none, we leave it as is. This corresponds to an error condition, and errors are "sticky." Otherwise, the value is of the form Option.some a, and we apply the operation f on a—or bind f's argument to a. We can now use connect to program our sum function:

```
def sum257Connect (ns : List \mathbb{N}) : Option \mathbb{N} :=
  connect (nth ns 1)
   (fun n_2 \mapsto \text{connect (nth ns 4)}
      (fun n_5 \mapsto \text{connect (nth ns 6)}
      (fun n_7 \mapsto \text{Option.some } (n_2 + n_5 + n_7))))
```

Intuitively, the program performs the following steps:

- 1. Extract the second item from the list using nth. If there is no such item, nth returns Option.none; simply return this value. Otherwise, bind n2 to this item and continue with the next step.
- 2. Perform the same for the fifth and seventh item, mutatis mutandis.
- 3. Return the sum of n2, n5, and n7 in an Option.some wrapper.

Mathematically, our new function sum257Connect is equal to the original function sum257.

Instead of defining connect ourselves, we could have used Lean's predefined general bind operation. It takes the same arguments in the same order. Here is the new code:

```
def sum257Bind (ns : List \mathbb{N}) : Option \mathbb{N} :=
  bind (nth ns 1)
   (fun n_2 \mapsto \text{bind (nth ns 4)}
        (fun n_5 \mapsto \text{bind (nth ns 6)}
        (fun n_7 \mapsto \text{pure } (n_2 + n_5 + n_7))))
```

We also use the predefined pure function instead of Option.some to convert a pure  $\alpha$  value to an Option  $\alpha$ .

One of the advantages of using the predefined bind is that it provides syntactic sugar, in the form of the >>= operator:

```
def sum2570p (ns : List \mathbb{N}) : Option \mathbb{N} := nth ns 1 >>=
    fun n_2 \mapsto nth ns 4 >>=
    fun n_5 \mapsto nth ns 6 >>=
    fun n_7 \mapsto pure (n_2 + n_5 + n_7)
```

The syntax oa >>= f expands to bind oa f, where oa is of type Option  $\alpha$ .

The next-to-last version of the sum program uses heavier syntactic sugar:

```
\begin{array}{l} \text{def sum257Dos (ns : List $\mathbb{N}$) : Option $\mathbb{N}$ :=} \\ \text{do} \\ \text{let $n_2$} \leftarrow \text{nth ns 1} \\ \text{do} \\ \text{let $n_5$} \leftarrow \text{nth ns 4} \\ \text{do} \\ \text{let $n_7$} \leftarrow \text{nth ns 6} \\ \text{pure ($n_2$ + $n_5$ + $n_7$)} \end{array}
```

The do notation provides a convenient syntax for effectful programs. The program do let  $a \leftarrow oa \dots$  is equivalent to  $oa >>= (fun a \mapsto \dots)$ . If we are not interested in the result of oa's computation, we can omit the let  $a \leftarrow binding$  and write do  $oa \dots$ , which expands to  $oa >>= (fun \_ \mapsto \dots)$ .

The do notation conveniently allows multiple let bindings in a single block. This brings us to the final version of the program:

```
def sum257Do (ns : List \mathbb{N}) : Option \mathbb{N} := do let n_2 \leftarrow nth ns 1 let n_5 \leftarrow nth ns 4 let n_7 \leftarrow nth ns 6 pure (n_2 + n_5 + n_7)
```

Each line with an arrow ← attempts to read a value. In case of failure, the entire program evaluates to Option.none.

The above function can be read as an imperative program where each of the nth calls can throw an exception. But even though the notation has an imperative flavor, the function is a pure functional program.

## 7.2 Two Operations and Three Laws

The Option type constructor is an example of a monad, called the *option monad*. In general, a monad is a unary type constructor  $m: \mathsf{Type} \to \mathsf{Type}$  that depends on some type parameter  $\alpha$  equipped with two distinguished operations:

```
pure \{\alpha: \mathsf{Type}\}: \alpha \to \mathsf{m} \ \alpha
bind \{\alpha \ \beta: \mathsf{Type}\}: \mathsf{m} \ \alpha \to (\alpha \to \mathsf{m} \ \beta) \to \mathsf{m} \ \beta
```

As usual, curly braces denote implicit arguments. For options, the pure operation is simply Option.some, whereas bind amounts to our connect.

A value of type m  $\alpha$  is an effectful program. The pure operation embeds a pure, effectless program of type  $\alpha$  in m  $\alpha$ . The bind operation composes two effectful programs, of types m  $\alpha$  and m  $\beta$ . The first program's output, of type  $\alpha$ , is passed to the second program. The second program's output is also the output of the composite program.

We can think of a monad as a box containing some data. The box captures some special effect (e.g., exceptions, a mutable state). The pure operation puts data into the box, whereas bind allows us to access the data in the box and modify it—possibly even changing its type, since the result has type m  $\beta$ , not m  $\alpha$ . There is, however, no general way to extract the data from the box—i.e., to obtain an  $\alpha$  from an m  $\alpha$ . There might not be any  $\alpha$  value in it, or there might be several.

To summarize, pure a is a box containing the value a, with no effects, whereas bind ma f (also written ma >>= f or do a  $\leftarrow$  ma, f a) executes ma, then executes f with the unboxed result a of ma. It is convenient to use names such as ma or mb for boxed values of type m  $\alpha$  or m  $\beta$ , and a or b for data of type  $\alpha$  or  $\beta$ .

Monads are an abstract concept with many applications. The option type is only one instance among many. The following table gives an overview of some monad instances and their effects.

Туре	Effect
id	no effects
Option	simple exceptions
fun $\alpha \mapsto \sigma \to \alpha \times \sigma$	threading through a state of type $\sigma$
Set	nondeterministic computations returning $\alpha$ values
$\text{fun }\alpha\mapsto t\to\alpha$	reading elements of type t (e.g., a configuration)
fun $\alpha\mapsto \mathbb{N}\times \alpha$	adjoining running time (e.g., to model time complexity)
$\text{fun }\alpha\mapsto \text{String}\times\alpha$	adjoining text output (e.g., for logging)
IO	interaction with the operating system
TacticM	interaction with the proof assistant

All of the above are unary type constructors  $m: Type \to Type$ . Some effects can be combined (e.g., fun  $\alpha \mapsto Option \ (t \to \alpha)$ ). Some effects are not executable (e.g., Set); these are nonetheless useful for modeling programs abstractly. Specific type constructors m may provide further operators beyond pure and bind. For example, they may provide a way to extract the boxed value.

Monads have several benefits. They provide the highly readable do notation. They support generic operations, such as List.mmap  $\{\alpha \ \beta : \mathsf{Type}\} : (\alpha \to \mathsf{m} \ \beta) \to \mathsf{List} \ \alpha \to \mathsf{m} \ (\mathsf{List} \ \beta)$ , that work uniformly for all monads m. To quote *Programming in Lean* [2]:

The power of the abstraction is not only that it provides general functions and notation that can be used in all these various instantiations, but also that it provides a helpful way of thinking about what they all have in common.

Besides being a useful computer science concept, monads provide a nice example of axiomatic reasoning.

The bind and pure operations are normally required to obey three laws. The bind operation combines two programs. If either of these is a pure program, we can inline it and eliminate the bind. This gives us the first two laws:

```
do let a' \leftarrow pure a = fa fa' and do let a \leftarrow ma = ma pure a
```

The third law is an associativity rule for bind. It allows us to flatten a nested computation:

7.3. A Type Class

Earlier we compared a monad to a box. It may help to think of the box more concretely as a Swiss bank account, with  $\alpha$ : = money. The first law means that if you put some money in the account, you can take it out. The second law means that nobody will notice if you take out some money and put it back afterwards. The third law means that performing two bank operations together followed by a third one is the same as performing the first one alone followed by the other two. Given the Swiss banks' reputation for secrecy, all three laws seem plausible.

#### 7.3 A Type Class

Monads are a mathematical structure, so we use a type class to specify them in Lean. Recall that a type class is a structure type that is parameterized, typically by a type, but here by a type constructor  $m: \mathsf{Type} \to \mathsf{Type}$ . Whenever we use a field from the type class on a concrete m, the type class inference mechanism retrieves the relevant structure value—the type class instance.

A possible Lean definition of monads, together with the three laws, follows:

```
class LawfulMonad (m : Type \rightarrow Type)
  extends Pure m, Bind m where
  pure_bind {\alpha \beta : Type} (a : \alpha) (f : \alpha \rightarrow m \beta) :
    (pure a >>= f) = f a
  bind_pure {\alpha : Type} (ma : m \alpha) :
    (ma >>= pure) = ma
  bind_assoc {\alpha \beta \gamma : Type} (f : \alpha \rightarrow m \beta) (g : \beta \rightarrow m \gamma)
    (ma : m \alpha) :
    ((ma >>= f) >>= g) = (ma >>= (fun a \mapsto f a >>= g))
```

Let us study this definition step by step:

- We are creating a structure parameterized by a unary type constructor m—i.e., a value of type Type → Type.
- The structure inherits the fields, and any syntactic sugar, from structures called Pure and Bind. These provide the pure and bind operations on m, with the expected types and the syntactic sugar.
- Finally, three fields (pure\_bind, bind\_pure, and bind\_assoc) are added to those already provided by Pure and Bind. Each field is a proof of one of the three laws.

We call our type class LawfulMonad because the three laws are required to hold. To instantiate this definition, we must supply the type constructor m, suitable bind and pure operators, and proofs of the laws.

Lean includes its own concept of monads, also called Lawful Monad. It is roughly equivalent to our definition but is distributed over multiple type classes.

#### 7.4 No Effects

Lean's constant id  $\{\alpha: \mathsf{Type}\}: \alpha \to \alpha$  is defined as the identity function fun  $\mathsf{x} \mapsto \mathsf{x}$ . The identity type constructor is obtained by taking  $\alpha:=\mathsf{Type}$ . We can register it as a monad:

```
def id.pure \{\alpha : \mathsf{Type}\} : \alpha \to \mathsf{id}\ \alpha
   | a => a
def id.bind \{\alpha \ \beta : \mathsf{Type}\} : \mathsf{id} \ \alpha \to (\alpha \to \mathsf{id} \ \beta) \to \mathsf{id} \ \beta
   | a, f => f a
instance id.LawfulMonad : LawfulMonad id :=
   { pure
                    := id.pure
      bind
                      := id.bind
      pure_bind :=
        by
            intro \alpha \beta a f
           rf1
      bind_pure :=
        by
            intro \alpha ma
            rfl
      bind_assoc :=
        by
            intro \alpha \beta \gamma f g ma
            rfl }
```

The registration process requires us to provide five components: the pure and bind operations and the proofs of the three laws.

The *identity monad* is the simplest monad possible. It provides a simple box, with a single value in it, without any effect. It plays a similar role as 0 in additive arithmetic. We can think of the other monads as variations of it; for example, the option monad is an identity monad enriched with a special Option.none value representing an error state.

## **7.5 Basic Exceptions**

As we saw above, the option type provides a basic exception mechanism. The following code shows how to register Option: Type  $\rightarrow$  Type as a lawful monad:

```
def Option.pure \{\alpha: \mathsf{Type}\}: \alpha \to \mathsf{Option} \ \alpha:= \mathsf{Option.some} def Option.bind \{\alpha \ \beta: \mathsf{Type}\}: \mathsf{Option} \ \alpha \to (\alpha \to \mathsf{Option} \ \beta) \to \mathsf{Option} \ \beta | \mathsf{Option.none}, \quad \_ => \mathsf{Option.none} | \mathsf{Option.some} \ a, \ f => f \ a instance Option.LawfulMonad : LawfulMonad Option :=
```

```
{ pure
           := Option.pure
  bind
             := Option.bind
  pure_bind :=
    by
      intro \alpha \beta a f
      rfl
  bind_pure :=
    by
      intro \alpha ma
      cases ma with
      | none => rfl
      | some _ => rfl
  bind_assoc :=
    by
      intro \alpha \beta \gamma f g ma
      cases ma with
      | none => rfl
      | some _ => rfl }
```

The three proofs are straightforward.

Beyond the standard operations, it can be useful to throw and catch exceptions. This can be implemented as follows:

```
def Option.throw \{\alpha: \mathsf{Type}\}: \mathsf{Option}\ \alpha:= \mathsf{Option.none} def Option.catch \{\alpha: \mathsf{Type}\}: \mathsf{Option}\ \alpha \to \mathsf{Option}\ \alpha \to \mathsf{Option}\ \alpha \to \mathsf{Option}\ \alpha | Option.none, ma' => ma' | Option.some a, _ => Option.some a
```

The Option.throw operation raises an exception, leaving the program in an error state (Option.none). The Option.catch operation can be used to recover from an earlier exception. If the program is currently in an error state, Option.catch invokes some exception-handling code (its second argument). This code might in turn raise a new exception. If Option.catch is applied to a normal state (of the form Option.some a), nothing happens.

As a convenient alternative to Option.catch ma ma', Lean supports the syntax ma.catch ma'. Here is a schematic example that demonstrates throwing and catching with this syntax:

```
do
...
if ... then
Option.throw
else
...
.catch do
```

The corresponding Java code would look as follows:

```
try {
```

```
if (...) {
    throw new UnknownException();
} else {
    ...
}
catch (UnknownException e) {
    ...
}
```

Options cater for only one kind of error state. A more general abstraction, called *error monad*, supports different errors, as with the exceptions of Java and other programming languages.

#### 7.6 Mutable State

The state monad provides an abstraction corresponding to a mutable state. For some programming languages, the compiler can detect the use of the state monad and translate programs using them to more efficient imperative programs.

Admittedly, "abstraction corresponding to a mutable state" may sound somewhat abstract, so let us consider a semiconcrete example. If you have some experience with functional programming, you probably at some point or other have written code that looks very much like this fragment:

```
def welcomeNewUser (userName : String) (ctxt : Context) :
  (N × String) × Context :=
  let
    (user, ctxt') := createUser userName ctxt
    (password, ctxt'') := generateTemporaryPassword user ctxt'
    (ok, ctxt''') := sendUnencryptedEmail user password ctxt''
in
    ((user, password), ctxt''')
```

(It is very important to send the password in an unencrypted email, and to ignore the ok status of the function.) The function takes some global state or context as input and invokes three functions in turn, each of which takes a context value and returns both some data and a new context. The context is effectively threaded through the program. This allows us to have a mutable state in a programming language without side effects.

The above approach is error-prone—it is all too easy to forget one prime (') and pass the wrong context to a function. The code is also cluttered by all the context variables. What if we could simply write the following instead?

```
def welcomeNewUserDo (userName : String) :
   Context \rightarrow (N \times String) \times Context :=
   do
   let user \leftarrow createUser userName
   let password \leftarrow generateTemporaryPassword user
   let ok \leftarrow sendUnencryptedEmail user password
   pure (user, password)
```

7.6. Mutable State

This is exactly what the state monad provides.

The state monad builds on top of a binary type constructor Action, which captures the concept of computations or actions over states of type  $\sigma$  with return values of type  $\alpha$ . In Lean, Action  $\sigma$   $\alpha$  is defined as equal to  $\sigma \to \alpha \times \sigma$ :

```
def Action (\sigma \alpha : Type) : Type := \sigma \rightarrow \alpha \times \sigma
```

(Since types are terms, we can use def also to define type abbreviations.) For a given type  $\sigma$ , we have that Action  $\sigma$ : Type  $\to$  Type is a monad. The type  $\sigma$  abstracts over the exact memory layout. We could use tuples or lists to represent memory, for example, and instantiate the abstract state  $\sigma$  accordingly.

A stateful action is a function that takes some state and returns both a value and some new state. The  $\sigma \to \text{part}$  of Action's definition gives the old state; the left component of the cartesian product,  $\alpha$ , gives the result of the computation; and the right component of the product,  $\sigma$ , gives the new state. Thus, the state is implicitly threaded through the program. As with other effectful programs, the do notation only exposes the data—the value of type  $\alpha$ —and conceals the effect—the old and new  $\sigma$  states.

The case where  $\sigma$  := Unit, a type of cardinality one (similar to void in C or Java) whose unique value is written (), corresponds to the identity monad: The type Unit  $\to \alpha \times$  Unit is isomorphic to  $\alpha$ . This intuition can guide us when defining pure and bind.

We start by defining basic operations: two operations, read and write, to access the memory, and the standard operations bind and pure:

```
def Action.read \{\sigma: \mathsf{Type}\}: Action \sigma \sigma | s => (s, s) def Action.write \{\sigma: \mathsf{Type}\} (s:\sigma): Action \sigma Unit | \_ => ((), s) def Action.pure \{\sigma \ \alpha: \mathsf{Type}\} (a:\alpha): Action \sigma \ \alpha | s => (a, s) def Action.bind \{\sigma: \mathsf{Type}\} \{\alpha \ \beta: \mathsf{Type}\} (\mathsf{ma}: \mathsf{Action} \ \sigma \ \alpha) (f:\alpha \to \mathsf{Action} \ \sigma \ \beta): Action \sigma \ \beta | s => match \ \mathsf{ma} \ s \ \mathsf{with} | (a, s') => f \ a \ s'
```

The read operation simply returns the current state s (in the first pair component) and leaves the state unchanged (in the second pair component). The write operation replaces the current state with s and returns (). The pure operation returns the current state s unchanged tupled together with the given value a. The bind operation passes the initial state to the ma argument, yielding a result a and a new state s'. These are passed to f, which returns a new result and a new state.

To register the Action type constructor as a lawful monad, we need as before to prove the three laws:

<sup>&</sup>lt;sup>1</sup>This type constructor is traditionally called State, but this name is very confusing.

```
instance Action.LawfulMonad \{\sigma : Type\}:
  LawfulMonad (Action \sigma) :=
  { pure
                := Action.pure
                 := Action.bind
    bind
    pure_bind :=
       by
         intro \alpha \beta a f
         rfl
    bind_pure :=
       by
         intro \alpha ma
         rfl
    bind_assoc :=
       by
         intro \alpha \beta \gamma f g ma
```

Notice that, in all three proofs, we unfold the definitions of the generic pure and bind constants to expand them to Action.pure and Action.bind, whose definitions we also expand.

As a concrete example, the following program removes all elements that are smaller than a previous element in the list, leaving us with a list of increasing elements. The maximal element is stored as the state  $\sigma$ . Notice how the state is accessed by read and write.

To execute the program, we must supply an initial state. The last state is returned along with the resulting list. It corresponds to the largest element encountered in the list or the start state. Thus, the commands

```
#eval increasingly [1, 2, 3, 2] 0
#eval increasingly [1, 2, 3, 2, 4, 5, 2] 0
```

produce the output

```
([1, 2, 3], 3)
([1, 2, 3, 4, 5], 5)
```

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#### 7.7 Nondeterminism

Whereas the option monad stores zero or one  $\alpha$  values and the identity and state monads store exactly one value, the *set monad* stores a possibly infinite number of values. This is useful to model nondeterminism, as a set of possible behaviors.

Lean's type Set  $\alpha$  is defined as  $\alpha \to \mathsf{Prop}$ . In other words, a set is identified with its characteristic predicate. Familiar operators such as the empty set ( $\emptyset$ ), the universal set (Set.univ), union ( $\cup$ ), intersection ( $\cap$ ), and membership ( $\in$ ) are supported, as well as traditional curly brace notations such as {a}, {a, b}, and {x | P x}. Many set constructs can be simplified by simp.

The Set type constructor can be registered as a lawful monad as follows:

```
def Set.pure \{\alpha : \mathsf{Type}\} : \alpha \to \mathsf{Set}\ \alpha
  | a => \{a\}
def Set.bind \{\alpha \ \beta : \mathsf{Type}\} : \mathsf{Set} \ \alpha \to (\alpha \to \mathsf{Set} \ \beta) \to \mathsf{Set} \ \beta
   | A, f \Rightarrow \{b \mid \exists a, a \in A \land b \in f a\}
instance Set.LawfulMonad : LawfulMonad Set :=
  { pure
                   := Set.pure
     bind
                   := Set.bind
     pure_bind :=
        by
           intro \alpha \beta a f
           simp [Pure.pure, Bind.bind, Set.pure, Set.bind]
     bind_pure :=
        by
           intro \alpha ma
           simp [Pure.pure, Bind.bind, Set.pure, Set.bind]
     bind_assoc :=
        by
           intro \alpha \beta \gamma f g ma
           simp [Pure.pure, Bind.bind, Set.pure, Set.bind]
           apply Set.ext
           aesop }
```

The pure operation simply puts the given value a in a singleton set  $\{a\}$ . The bind operation calls f on all values in the set A and returns the union of all the results. For example, if  $A := \{3, 8\}$  and  $f := \{a + 1, a + 2\}$ , then set.bind A f equals  $\{4, 5, 9, 10\}$ .

The last proof relies on set extensionality, which states that two sets that contain the same elements must be equal:

```
Set.ext \{\alpha : \mathsf{Type}\}\ \{\mathsf{A}\ \mathsf{B} : \mathsf{set}\ \alpha\} : (\forall \mathsf{x},\,\mathsf{x} \in \mathsf{A} \leftrightarrow \mathsf{x} \in \mathsf{B}) \to \mathsf{A} = \mathsf{B}
```

Another noteworthy point is the use of the aesop tactic. The goal's target is

```
\forall x, x \in b \mid \exists a, (\exists a\_1, a\_1 \in ma \land a \in f a\_1) \land b \in g a
 \leftrightarrow x \in b \mid \exists a, a \in ma \land \exists a\_1, a\_1 \in f a \land b \in g a\_1
```

where the two sides of  $\leftrightarrow$  are the same except for the placement of the existential quantifiers (and, confusingly, the names of the bound variables). This ugly propo-

sition could be proved by a tedious sequence of introduction and elimination, but we deserve more automation.

## 7.8 Aesop Tactic

#### aesop

The aesop tactic [19], whose name stands for "automated extensible search for obvious proofs," is a general-purpose proof search tactic. Among others, it performs elimination of the logical symbols  $\land$ ,  $\lor$ ,  $\leftrightarrow$ , and  $\exists$  in hypotheses and introduction of  $\land$ ,  $\leftrightarrow$ , and  $\exists$  in the target, and it regularly invokes simp. It can succeed at proving a goal, fail, or succeed partially, leaving some unfinished subgoals to the user.

## 7.9 A Generic Algorithm: Iteration over a List

Suppose we apply an effectful function f on all elements of a list using map. We then obtain a regular list of effectful values. For example:

```
def nthsFine \{\alpha : \mathsf{Type}\}\ (\mathsf{xss} : \mathsf{List}\ (\mathsf{List}\ \alpha))\ (\mathsf{n} : \mathbb{N}) :
List (Option \alpha) :=
List.map (fun \mathsf{xs} \mapsto \mathsf{nth}\ \mathsf{xs}\ \mathsf{n})\ \mathsf{xss}
```

The function nthsFine xss n tries to extract the (n + 1)st element of each list in xss. Running

```
#eval nthsFine [[11, 12, 13, 14], [21, 22, 23]] 2
returns [Option.some 13, Option.some 23].
```

These Option.some constructors can be inconvenient. Often, we only care about whether any error arose. This leads us to our final example: a generic effectful program mmap that iterates over a list and applies an effectful function f to each element. The definition is recursive:

```
def mmap \{m: Type \rightarrow Type\} [LawfulMonad m] \{\alpha \ \beta : Type\} (f : \alpha \rightarrow m \ \beta) :

List \alpha \rightarrow m (List \beta)

| [] => pure []

| a :: as => do

let b \leftarrow f a

let bs \leftarrow mmap f as

pure (b :: bs)
```

Notice that the function returns a single m value containing a list and not a list of m values. Try to work out why it is well typed and has the desired behavior.

We can now trying using mmap instead of map:

```
def nthsCoarse \{\alpha: \mathsf{Type}\}\ (\mathsf{xss}: \mathsf{List}\ (\mathsf{List}\ \alpha))\ (\mathsf{n}: \mathbb{N}):

Option (List \alpha) :=

mmap (fun \mathsf{xs}\mapsto \mathsf{nth}\ \mathsf{xs}\ \mathsf{n})\ \mathsf{xss}

Running

#eval nthsCoarse [[11, 12, 13, 14], [21, 22, 23]] 2
```

returns Option.some [13, 23], with a single Option.some around a pure list.

The mmap function distributes over the append operator ++. The do notation is useful not only for defining functions but also for stating their properties:

```
theorem mmap_append {m : Type \rightarrow Type} [LawfulMonad m] {\alpha \beta : Type} (f : \alpha \rightarrow m \beta) : \forallas as' : List \alpha, mmap f (as ++ as') = do let bs \leftarrow mmap f as let bs' \leftarrow mmap f as' pure (bs ++ bs') | [], _ => by simp [mmap, LawfulMonad.bind_pure, LawfulMonad.pure_bind] | a :: as, as' => by simp [mmap, mmap_append _ as as', LawfulMonad.pure_bind, LawfulMonad.bind_assoc]
```

### 7.10 Summary of Newly Introduced Lean Constructs

#### **Notations**

do indicates the start of an effectful program
let ... ← ... assigns a variable in an effectful program
>>= composes effectful computations

#### Theorem

Set.ext set extensionality

#### **Tactic**

aesop proves propositions using a generic search procedure

## **Chapter 8**

# Metaprogramming

Like most proof assistants, Lean can be extended with custom tactics and other functionality. Programming Lean itself, as opposed to merely using it, is called metaprogramming. Lean's metaprogramming framework mostly uses the same notions and syntax as Lean's input language, so that we do not need to learn a different language to program Lean. Monads are used to access Lean's state.

Abstract syntax trees, presented as inductive types, reflect internal data structures. The proof assistant's internals are exposed through Lean functions, which we can use to access the current goal, unify terms, query and modify the global context, and set attributes (e.g., @[simp]).

Here are some example applications of metaprogramming:

- goal transformations (e.g., applying safe introduction rules, putting the goal in negation normal form);
- heuristic proof search (e.g., applying unsafe introduction rules with backtracking);
- decision procedures (e.g., for linear arithmetic, propositional logic);
- definition generators (e.g., Haskell-style deriving for inductive types);
- advisor tools (e.g., theorem finders, counterexample generators);
- exporters (e.g., documentation generators);
- ad hoc proof automation (to avoid boilerplate or duplication).

As mathematician and Lean user Kevin Buzzard wrote:1

If you find yourself "grinding" (to use a computer game phrase), doing the same sort of stuff over and over again because you need to do it to make progress, then you can try to persuade a computer scientist to write a tactic to do it for you (or even write your own tactic if you're brave enough to write meta Lean code).

#### 8.1 Tactic Combinators

First, some terminology: A tactic *fails* if applying it produces an error; otherwise, it *succeeds*. One way for a tactic to succeed is to fully prove the goal. Another is

 $<sup>^{1}</sup>$ https://xenaproject.wordpress.com/2020/02/09/lean-is-better-for-proper-maths-than-all-the-other-theorem-provers/

to produce new subgoals that replace the current goal. Some tactics succeed by doing nothing.

When programming our own tactics, we often need to repeat some actions on several goals, or to recover if a tactic fails. Tactic combinators help in such cases. One of the most useful tactic combinators is repeat' tactic. It invokes tactic repeatedly on all goals, then on the emerging subgoals, then on the emerging subsubgoals, and so on until tactic fails on all the available goals. Here is an example of repeat' involving the Even predicate presented in Chapter 6:

```
theorem repeat'_example :
   Even 4 ∧ Even 7 ∧ Even 3 ∧ Even 0 :=
   by
    repeat' apply And.intro
   repeat' apply Even.add_two
```

After the first repeat' line, the proof state consists of four goals:

Notice that all conjunctions are gone. The second repeat', which applies the theorem Even.add\_two:  $\forall k$ , Even  $k \rightarrow$  Even (k + 2) over and over, leaves us with these goals:

```
⊢ Even 0 ⊢ Even 1 ⊢ Even 1 ⊢ Even 0
```

The first and last goals are annoying because they correspond to the theorem Even.zero. We can prove them by trying to apply Even.zero whenever applying Even.add\_two fails. This is achieved as follows:

```
theorem repeat'_first_example :
   Even 4 \( \) Even 7 \( \) Even 3 \( \) Even 0 :=
   by
    repeat' apply And.intro
   repeat'
    first
        | apply Even.add_two
        | apply Even.zero
```

The tactic combinator first |  $tactic_1$  |  $\cdots$  |  $tactic_n$  first tries to execute its first argument,  $tactic_1$ . If this fails,  $tactic_2$  is tried, and so on. If all the specified tactics fail, the entire combinator fails. In the example above, we have two unprovable goals left:

```
⊢ Even 1 ⊢ Even 1
```

The next combinator, all\_goals tactic, invokes a tactic exactly once on each goal. The combinator succeeds only if tactic succeeds on all goals. It fails in the example below, because Even.add\_two cannot be applied to the goal  $\vdash$  Even o:

```
theorem all_goals_example :
   Even 4 \wedge Even 7 \wedge Even 3 \wedge Even 0 :=
   by
    repeat' apply And.intro
   all_goals apply Even.add_two -- fails
```

To ignore failures of tactic, we can wrap it in the try combinator:

```
theorem all_goals_try_example :
    Even 4 \( \) Even 7 \( \) Even 3 \( \) Even 0 :=
    by
        repeat' apply And.intro
        all_goals try apply Even.add_two

The resulting state is
    \( \) Even 2 \( \) Even 5 \( \) Even 1 \( \) Even 0
```

The construct try tactic is equivalent to first | tactic | skip, where skip is a tactic that succeeds without doing anything. Hence try tactic always succeeds. A related tactic is done: It succeeds if there are no goals left and fails otherwise.

Another variant is the any\_goals tactic combinator. It tries to invoke tactic once on each goal, but unlike all\_goals, it succeeds if tactic succeeds on any goal. The example

```
theorem any_goals_example :
    Even 4 \( \) Even 7 \( \) Even 3 \( \) Even 0 :=
    by
        repeat' apply And.intro
        any_goals apply Even.add_two

results in the state
    \( \) Even 2 \( \) Even 5 \( \) Even 1 \( \) Even 0
```

This is the same state as in the previous example. In general, the difference is that any goals tactic can fail whereas all goals try tactic always succeeds.

Sometimes we want to leave a goal alone unless we can fully prove it. The combinator solve  $|tactic_1| \cdots |tactic_n|$  first tries to execute its first argument,  $tactic_1$ . If this fails to prove the goal,  $tactic_2$  is tried, and so on. If all the specified tactics fail to prove the goal, the entire combinator fails. (Compare this behavior with that of first  $|tactic_1| \cdots |tactic_n|$ , which only requires one of the specified tactics to succeed, not to prove the goal.) Consider this example:

The first and fourth goals are proved, and we are left with the two unprovable goals exactly as they stand in the theorem statement:

```
⊢ Even 7 ⊢ Even 3
```

Note that the repeat' combinator can lead to infinite looping. Consider this example:

```
theorem repeat'_Not_example :
    ¬ Even 1 :=
    by repeat' apply Not.intro
```

The Not.intro rule is  $(?a \rightarrow False) \rightarrow \neg$  ?a, so it applies once to transform the goal into  $\vdash$  Even 1  $\rightarrow$  False. Because  $\neg$  ?a is defined as ?a  $\rightarrow$  False, the rule applies again, yielding the same goal again. The tactic loops.

Finally, the "and then" operator <; > can be used to join two tactics. The lefthand side is executed on the first goal. The right-hand side is executed on every emerging subgoal (but not on the original second goal, third goal, etc.). Thus, we can write

```
induction n <;>
aesop
```

instead of the more verbose

#### 8.2 Macros

Let us do some actual metaprogramming, by coding a custom tactic as a macro. The tactic embodies the behavior we hardcoded in the solve example above:

```
macro "intro_and_even" : tactic =>
  '(tactic|
        (repeat' apply And.intro
        any_goals
        solve
        | repeat'
        first
        | apply Even.add_two
        | apply Even.zero))
```

The first line declares <code>intro\_and\_even</code> as a macro belonging to the <code>tactic</code> syntactic category. On the remaining lines, the '(tactic| tactic) construct embeds the specified tactic, tactic, in a macro. The tactic itself is specified using standard syntax.

Once we have defined a custom tactic, we can invoke it in a proof:

```
theorem intro_and_even_example :
   Even 4 ∧ Even 7 ∧ Even 3 ∧ Even 0 :=
   by
   intro_and_even
```

This yields the subgoals

```
⊢ Even 7 ⊢ Even 3
```

#### 8.3 The Metaprogramming Monads

Macros are a mechanism that can be used to program simple proof automation. For most metaprogramming tasks, however, we need to use the metaprogramming monads, MetaM and TacticM.

MetaM combines the attributes of several kinds of monads:

- It is a state monad providing access to the global context (including all definitions and inductive types), notations, and attributes (e.g., the list of @[simp] theorems), among others.
- It behaves like an option monad. The metaprogram failure signals that a tactic has failed.
- It supports tracing, so we can use the program logInfo to display messages.
- Like other monads, it supports imperative constructs such as for-in loops, continue statements, and return statements.

TacticM extends MetaM with goal management: It provides access to the list of goals. It also allows us to run Lean to fill in the implicit { } and type class [ ] arguments in expressions, to expand macros, and more.

Inside Lean, each goal is represented as a metavariable ?m standing for a missing term (typically, a proof term). Each metavariable has a type (typically, a proposition) and a local context specifying the variables and hypotheses that can be used to prove the goal associated with the metavariable.

Let us put the metaprogramming monad to some use by defining a tactic that uses TacticM's tracing facilities to display the Lean version number, the list of goals, and the target of the first goal (which Lean calls the *main* goal):

```
def traceGoals : TacticM Unit :=
   do
     logInfo m!"Lean version {Lean.versionString}"
     logInfo "All goals:"
     let goals ← getUnsolvedGoals
     logInfo m!"{goals}"
     match goals with
     | [] => return
     | _ :: _ =>
        logInfo "First goal's target:"
     let target ← getMainTarget
        logInfo m!"{target}"

elab "trace_goals" : tactic =>
        traceGoals
```

The code has many interesting features:

- The first line declares a function traceGoals of type TacticM Unit—the type of tactics that return Unit (a trivial type of cardinality one). Notice that a metaprogramming function is defined using the same syntax as any Lean function.
- The second line enters the monad. The remaining lines are effectful operations that access Lean's internals.

- The m!"..." syntax specifies a string in which every occurrence of { term}, where term is a Lean term, are evaluated and serialized into a string. For example, if Lean.versionString is "v4.6.0-rc1", then m! "Lean version {Lean.versionString}" evaluates to the string "Lean version v4.6.0-rc1".
- The last two lines register a new tactic called trace\_goal that simply calls the traceGoals function.

Here is an example of how we can use the new tactic:

```
theorem Even_18_and_Even_20 (\alpha : Type) (a : \alpha) : Even 18 \wedge Even 20 := by apply And.intro trace_goals intro_and_even
```

The output, which is made visible by hovering over trace\_goals, is as follows:

```
Lean version v4.6.0-rc1 All goals: [case left \alpha: Type a: \alpha \vdash Even 18, case right \alpha: Type a: \alpha \vdash Even 20] First goal's target: Even 18
```

Although Lean displays goals using the familiar goal syntax  $C \vdash P$ , they are actually metavariables.

The constants used in the above program have the following types:

```
logInfo : MessageData → TacticM Unit
getUnsolvedGoals : TacticM (List MVarId)
getMainTarget : TacticM Expr
```

where MessageData represents a message, MVarId represents a metavariable identifier, and Expr represents a term.

## 8.4 First Example: An Assumption Tactic

Our first larger example implements a hypothesis tactic that, like the predefined assumption tactic, looks for a hypothesis of the right type (i.e., the right proposition) and applies it to prove the goal:

```
def hypothesis : TacticM Unit :=
  withMainContext
   (do
        let target ← getMainTarget
```

In the hypothesis function, we start by extracting the first goal's target and the local context. To make sure that getLCtx gives us the local context of the current first goal, we pass the entire do block to the withMainContext function. In general, any TacticM computation is performed within an ambient local context that gives meaning to the free variables occurring in expressions. The withMainContext function sets this local context to the local context of the current first goal.

Inside the do block, we iterate over all declarations in the local context, using the convenient monadic construct for-in. For each local variable or hypothesis h that is not a so-called implementation detail (i.e., a hypothesis inserted by Lean that is invisible to users), we check if its type (typically, its proposition) is equal to the target up to computation and instantiation of metavariables, and if so, we get the metavariable ?m associated with the first goal and assign ?m := h, thereby proving the goal. Finally, we return.

Since goals are represented by metavariables, assigning a term to a metavariable ?m is Lean's low-level way of proving a goal. New metavariables occurring in that term correspond to new subgoals that must be proved.

A simple invocation of hypothesis follows:

```
theorem hypothesis_example \{\alpha: Type\} \{p: \alpha \to Prop\} \{a: \alpha\} (hpa: pa): pa:= by hypothesis
```

If we add tracing, we can see that  $\alpha$ , p, a, and hpa are tried in turn before the matching hypothesis hpa is found and successfully applied.

The example used the following new constants:

```
\label{eq:getlCtx} \texttt{getLCtx}: \texttt{TacticM LocalContext} \\ \texttt{LocalDecl.isImplementationDetail}: \texttt{LocalDecl} \rightarrow \texttt{Bool} \\ \texttt{isDefEq}: \texttt{Expr} \rightarrow \texttt{Expr} \rightarrow \texttt{TacticM Bool} \\ \texttt{LocalDecl.type}: \texttt{LocalDecl} \rightarrow \texttt{Expr} \\ \texttt{getMainGoal}: \texttt{TacticM MVarId} \\ \texttt{MVarId.assign}: \texttt{MVarId} \rightarrow \texttt{Expr} \rightarrow \texttt{TacticM Unit} \\ \texttt{LocalDecl.toExpr}: \texttt{LocalDecl} \rightarrow \texttt{Expr} \\ \texttt{failure} \left\{\alpha: \texttt{Type}\right\}: \texttt{TacticM} \alpha \\ \end{aligned}
```

## 8.5 Expressions

The metaprogramming framework revolves around the type Expr of expressions or terms. An important component of expressions are names, of type Name. We begin with names.

Names can be specified using a single backtick. For example, 'x represents the name x, which can be given to a variable or constant. When referring to constants, we must specify the full name, including namespaces; thus, to refer to the Even predicate of Chapter 6, we must write 'LoVe. Even and not 'Even.

If we want to refer to an existing constant, Lean offers the double-backtick syntax, which looks up the name using Lean's usual name elaboration rules and expands it to its full name. Thus, both ''Even and ''LoVe.Even refer to the name LoVe.Even, and if we write some name that is not declared, such as ''EvenIf, Lean gives an error.

The type Expr is defined as follows:

```
inductive Expr : Type where
  \mid const \mid : Name \rightarrow List Level \rightarrow Expr
  \mid sort : Level \rightarrow Expr
  | fvar : FVarId \rightarrow Expr
  | mvar : MVarId \rightarrow Expr
  | app
              : Expr 
ightarrow Expr 
ightarrow Expr
  | lam
              : Name 	o Expr 	o Expr 	o BinderInfo 	o Expr
  | bvar : Nat \rightarrow Expr
  | forallE : Name 
ightarrow Expr 
ightarrow Expr 
ightarrow BinderInfo 
ightarrow Expr
  | letE \,: Name 
ightarrow Expr 
ightarrow Expr 
ightarrow Expr 
ightarrow Bool 
ightarrow Expr
  | lit
              : Literal \rightarrow Expr
  | mdata : MData \to Expr \to Expr
  | proj | : Name \rightarrow Nat \rightarrow Expr \rightarrow Expr
```

Let us review the main constructors:

- Expr.const name levels represents a constant called name, such as Nat.add or N. The levels argument represents universe levels, a concept that will be explained in Chapter 12. For example, Expr.const ''Nat.add [] represents Nat.add and Expr.const ''Nat [] represents Nat (i.e., N).
- Expr.sort level is used to represent the types of types. For example, Expr. sort Level.zero stands for Prop, and Expr.sort (Level.succ Level.zero) stands for Type.
- Expr.fvar id represents a free variable in the local context (e.g., a, h). The id argument is a unique identifier for the variable.
- Expr.mvar id represents a metavariable, i.e., a variable ?m with a question mark. The id argument is a unique identifier for the metavariable.
- Expr.app t u represents the application of a function t to an argument u. For example, Expr.app (Expr.const ''Nat.succ [] (Expr.const ''Nat.zero []) represents Nat.succ Nat.zero.
- **Expr.** lam name  $\sigma$  t bi represents an anonymous function (or  $\lambda$ -expression). The name argument is the bound variable's name, the  $\sigma$  argument is the bound variable's type, the t argument is the function's body, and the bi

argument stores whether the variable is an explicit ( ), implicit { }, or type class [ ] argument.

Expr.bvar i represents a bound variable, using a notation known as De Bruijn index. Expr.var o refers to the variable bound by the closest binder, Expr.var 1 refers to the variable bound by the second closest binder, and so on. Thus,

represents fun x y :  $\mathbb{N} \mapsto x$ .

■ Expr.forallE name  $\sigma \tau$  bi represents a possibly dependent function type. The name argument is the name of the bound variable, the  $\sigma$  argument is the domain type, the  $\tau$  argument is the result type, and bi is as for Expr.lam above. For example,

```
\begin{split} & \text{Expr.forallE 'n (Expr.const ''Nat [])} \\ & (\text{Expr.app (Expr.const ''Even []) (Expr.bvar 0))} \\ & \text{BinderInfo.default} \end{split} & \text{represents (n : } \mathbb{N}) \rightarrow \text{Even n (also written } \forall \text{n : } \mathbb{N}, \text{ Even n), and} \\ & \text{Expr.forallE 'dummy (Expr.const 'Nat [])} \\ & (\text{Expr.const 'Bool []) BinderInfo.default} \end{split} & \text{represents } \mathbb{N} \rightarrow \text{Bool.}
```

## 8.6 Second Example: A Conjunction-Destructing Tactic

In this and the next section, we define two further tactics that accomplish well-defined tasks. The first of these two tactics, called destruct\_and, automates the elimination of conjunctions in premises. Our aim is to automate proofs such as the following:

```
theorem abc_a (a b c : Prop) (h : a \land b \land c) :
    a :=
    And.left h

theorem abc_b (a b c : Prop) (h : a \land b \land c) :
    b :=
    And.left (And.right h)

theorem abc_bc (a b c : Prop) (h : a \land b \land c) :
    b \land c :=
```

```
And.right h
theorem abc_c (a b c : Prop) (h : a \land b \land c) :
    c :=
    And.right (And.right h)
```

In each case, we would like to simply write by destruct\_and h as the proof.

Our tactic relies on a helper function, which takes as argument a proof term hP (initially the hypothesis h) from which we extract conjuncts:

```
partial def destructAndExpr (hP : Expr) : TacticM Bool :=
  withMainContext
    (do
       let target ← getMainTarget
       let P ← inferType hP
        \textbf{let} \ \textbf{eq} \ \leftarrow \ \textbf{isDefEq} \ \textbf{P} \ \textbf{target}
        if eq then
          let goal ← getMainGoal
          MVarId.assign goal hP
          return true
        else
          match Expr.and? P with
          | Option.none
                                  => return false
          | Option.some (Q, R) =>
            let hQ ← mkAppM ''And.left #[hP]
            let success ← destructAndExpr hQ
            if success then
              return true
            else
              let hR ← mkAppM ''And.right #[hP]
              destructAndExpr hR)
```

As with hypothesis, we pass the entire do block to the withMainContext function. This ensures that inferType and isDefEq operate within the right local context. Inside the do block, we first extract the first goal's target and hP's type (typically, its proposition) P. If they are equal up to computation and instantiation of metavariables, we close the goal by assigning to its metavariable, as we did in the hypothesis example, and we return true to indicate success. Otherwise, we check whether hP's proposition is of the form  $Q \land R$ . If so, we call the helper function recursively with the proof term  $hQ := And.left\ hP$ , which is a proof of Q. If this succeeds, we are done; otherwise, we try the proof term  $hR := And.right\ hP$ , which is a proof of R.

Note the presence of the keyword partial at the beginning of the function definition. It is needed here because Lean failed to prove that the function always terminates. Since the function will only be used as a metaprogram, and not inside a proposition, termination is optional, and we can disable the termination check by specifying partial.

Also noteworthy is the mkAppM function, which is used to construct a curried application of a constant to an array of arguments. Arrays are similar to lists but are written prefixed with the symbol # (e.g., #[1, 2, 3]). Using mkAppM is more convenient than applying the Expr.app constructor multiple times. Additionally, mkAppM

allows us to omit implicit arguments such as the propositions Q and R, which we would otherwise have to provide as arguments to And.left and And.right.

The main function has very little left to do:

```
def destructAnd (name : Name) : TacticM Unit :=
  withMainContext
   (do
      let h ← getFVarFromUserName name
      let success ← destructAndExpr h
      if ! success then
         failure)

elab "destruct_and" h:ident : tactic =>
  destructAnd (getId h)
```

The function retrieves the hypothesis h using getFVarFromUserName and calls the helper function. If the helper returns false, the tactic fails.

We can now use our new gadget on the motivating examples:

```
theorem abc_a_again (a b c : Prop) (h : a \lambda b \lambda c) :
    a :=
    by destruct_and h

theorem abc_b_again (a b c : Prop) (h : a \lambda b \lambda c) :
    b :=
    by destruct_and h

theorem abc_bc_again (a b c : Prop) (h : a \lambda b \lambda c) :
    b \lambda c :=
    by destruct_and h

theorem abc_c_again (a b c : Prop) (h : a \lambda b \lambda c) :
    c :=
    by destruct_and h
```

The following new constants were used in the above metaprogram:

```
\begin{array}{c} \text{inferType}: \texttt{Expr} \rightarrow \texttt{TacticM} \ \texttt{Expr} \\ \texttt{Expr.and?}: \texttt{Expr} \rightarrow \texttt{Option} \ (\texttt{Expr} \times \texttt{Expr}) \\ \texttt{mkAppM}: \texttt{Name} \rightarrow \texttt{Array} \ \texttt{Expr} \rightarrow \texttt{TacticM} \ \texttt{Expr} \\ \texttt{getFVarFromUserName}: \texttt{Name} \rightarrow \texttt{TacticM} \ \texttt{Expr} \end{array}
```

## 8.7 Third Example: A Direct Proof Finder

Sometimes we state a theorem, prove it, and later realize that the theorem already exists. This can be prevented by using prove\_direct, a tactic that traverses all available theorems and checks whether one of them can prove the current goal. We will review its code in steps.

The first step is a function isTheorem that returns true if a declaration is an axiom or a theorem and false otherwise:

```
def isTheorem : ConstantInfo → Bool
    | ConstantInfo.axiomInfo _ => true
    | ConstantInfo.thmInfo _ => true
    | _ => false
```

We will use this function to filter out the declarations that do not interest us. The next function applies the theorem called name to the current goal:

```
def applyConstant (name : Name) : TacticM Unit :=
  do
    let cst ← mkConstWithFreshMVarLevels name
    liftMetaTactic (fun goal → MVarId.apply goal cst)
```

Given a name, the mkConstWithFreshMVarLevels function creates an expression cst representing the constant. The allusion to "fresh metavariable levels" in the function's name will become clearer in Chapter 12. MVarId.apply (not to be confused with MVarId.assign) then applies the constant to the current goal, setting ?m := cst ?m<sub>1</sub> ... ?m<sub>n</sub> and returning the fresh metavariables ?m<sub>j</sub>, which represent the premises of cst.

The liftMetaTactic function retrieves the identifier of the first goal, runs the given function on the goal within the lower-level MetaM monad, and replaces the goal with the subgoals returned by the function.

The next function implements a combinator that behaves like <; > but that can be used from a metaprogram:

```
def andThenOnSubgoals (tac<sub>1</sub> tac<sub>2</sub> : TacticM Unit) :
  TacticM Unit :=
  do
    let origGoals ← getGoals
    let mainGoal ← getMainGoal
    setGoals [mainGoal]
    let subgoals_1 \leftarrow getUnsolvedGoals
    let mut newGoals := []
    for subgoal in subgoals, do
       let assigned \leftarrow MVarId.isAssigned subgoal
       if ! assigned then
         setGoals [subgoal]
         tac<sub>2</sub>
         \textbf{let} \ \textbf{subgoals}_2 \ \leftarrow \ \textbf{getUnsolvedGoals}
          newGoals := newGoals ++ subgoals<sub>2</sub>
    setGoals (newGoals ++ List.tail origGoals)
```

The TacticM monad keeps track of the current goals to prove. We can retrieve the list using getGoals and set it using setGoals. Setting the list of subgoals is useful if we want the tactic to temporarily focus on specific subgoals.

Here, we first focus on the first goal (setGoals [mainGoal]) and invoke the first tactic. For each subgoal that emerges, we focus on it (setGoals [subgoal]) and invoke the second tactic. All the unsolved subsubgoals emerging from the second tactic are collected in the mutable variable newGoals. Since proving a goal can sometimes instantiate another metavariable, we check at each iteration whether the current subgoal metavariable is assigned and skip the subgoal if it is.

At the end, we update the goals to include all the pending goals: those in newGoals and all but the first goal of origGoals, which we have not considered.

In general, at the end of a tactic, we should make sure that the list of goals consists of all goals that remain to be proved. Otherwise, we may get cryptic errors such as

```
declaration has metavariables
```

We also need a tactic that attempts to prove the goal using a theorem specified by its name and that invokes hypothesis to prove any emerging subgoals:

```
def proveUsingTheorem (name : Name) : TacticM Unit :=
  andThenOnSubgoals (applyConstant name) hypothesis
```

This is the programmatic equivalent of the proof apply name <; > hypothesis. Finally, we are ready to review the main function:

```
def proveDirect : TacticM Unit :=
    let origGoals \leftarrow getUnsolvedGoals
    let goal ← getMainGoal
    setGoals [goal]
    let env ← getEnv
    for (name, info)
        in SMap.toList (Environment.constants env) do
      if isTheorem info && ! ConstantInfo.isUnsafe info then
          proveUsingTheorem name
          logInfo m!"Proved directly by {name}"
          setGoals (List.tail origGoals)
          return
        catch _ =>
          continue
    failure
elab "prove_direct" : tactic =>
  proveDirect
```

We focus on the first goal, then iterate over all constants declared in the environment. If the constant is a theorem and is not so-called unsafe, we try to apply it using our helper proveUsingTheorem. If this succeeds, we print "Proved directly by name," where name is the name of the theorem, and return. On failure, we keep iterating. If the entire iteration is exhausted, we report a failure.

Here is the tactic in action:

```
theorem Nat.symm (x y : N) (h : x = y) :
  y = x :=
  by prove_direct
```

This prints "Proved directly by symm." The message is helpful because we can apply the specified theorem directly instead of relying on the relatively slow prove\_direct tactic. Specifically, we can apply the theorem symm in conjunction with hypothesis as follows:

```
theorem Nat.symm_manual (x y : N) (h : x = y) :
    y = x :=
    by
    apply symm
    hypothesis
```

Here is a list of new constants featured in this example:

```
mkConstWithFreshMVarLevels : Name → TacticM Expr

liftMetaTactic : (MVarId → MetaM (List MVarId)) →

TacticM Unit

MVarId.apply : MVarId → Expr → MetaM (List MVarId)

getGoals : TacticM (List MVarId)

setGoals : List MVarId → TacticM Unit

MVarId.isAssigned : MVarId → TacticM Bool

getEnv : TacticM Environment

SMap.toList : ConstMap → List (Name × ConstantInfo)

Environment.constants : Environment → ConstMap

ConstantInfo.isUnsafe : ConstantInfo → Bool
```

This concludes our review of prove\_direct. A similar, more sophisticated tactic is available as apply? in mathlib.

#### 8.8 Miscellaneous Tactics

Although the focus of this chapter was on developing new tactics, we encountered three predefined tactics.

#### skip

The skip tactic succeeds without doing anything. It is sometimes useful as a building block when we develop custom tactics.

#### done

The done tactic raises a failure if there are some goals left; otherwise, it succeeds without doing anything. Like skip, it can be useful as a building block.

#### apply?

The apply? tactic searches the loaded libraries for a lemma that exactly proves the goal. On success, it suggests a tactic invocation of the form exact ..., which can be inserted in the formalization.

## 8.9 Summary of Newly Introduced Lean Constructs

#### **Declaration**

partial

prefixes declarations of possibly nonterminating metaprograms

## Quotations

`n	quotes a literal name
``n	quotes a literal name with elaboration and checking

## **Tactics**

apply?	searches for a theorem that proves the current goal
done	fails if there are some goals left
skip	does nothing

## **Tactic Combinators**

<;>	invokes the second tactic on all subgoals from the first tactic
all_goals	invokes a tactic once on each goal, expecting only successes
any_goals	invokes a tactic once on each goal, expecting at least one success
first   · · ·   · · ·	tries the tactics in turn until one succeeds
repeat'	repeatedly invokes a tactic on all goals and subgoals until failure
solve   ···   ···	tries to fully prove the current goal using the tactics in turn
try	tries to invoke a tactic; does nothing on failure

# Part III Program Semantics

## **Chapter 9**

## **Operational Semantics**

In this and the next two chapters, we will see how to use Lean to specify the syntax and semantics of programming languages, to prove properties of the semantics, and to reason about concrete programs.

This chapter is heavily inspired by Chapter 7 of Concrete Semantics: With Isabelle/HOL [23].

#### 9.1 Formal Semantics

A formal semantics allows us to specify and reason about a programming language and about individual programs written in that language. It can form the basis of verified compilers, interpreters, verifiers, static analyzers, type checkers, and more. Without formal proofs, these tools are almost always incorrect.

Consider WebAssembly, a new assembly-like language for web browsers, designed as a portable target for compiling high-level languages such as C++ and Rust. A researcher, Conrad Watt [28], formalized its semantics and type system using the Isabelle/HOL proof assistant. He found many issues (our italics):

We have produced a full Isabelle mechanisation of the core execution semantics and type system of the WebAssembly language. In addition, we have created a mechanised proof for the type soundness properties stated in the working group's paper. In order to complete this proof, several deficiencies in the official WebAssembly specification, uncovered by our proof and modelling work, needed to be corrected by the specification authors. In some cases, these meant that the type system was originally unsound.

We have maintained a constructive dialogue with elements of the working group, mechanising and verifying new features as they are added to the specification. In particular, the mechanism by which a Web-Assembly implementation interfaces with its host environment was not formally specified in the working group's original paper. Extending our mechanisation to model this feature revealed a deficiency in the WebAssembly specification that sabotaged the soundness of the type system.

Watt's research is only one example among many. Proof assistants are widely used for programming language research. Every year, around 10%–20% of papers

presented at the Principles of Programming Languages (POPL) conference are formalized. This is possible because comparatively little machinery is needed to get started. The proofs tend to have lots of cases, which is a good match for computers. Moreover, proof assistants are extremely convenient to keep track of what needs to be changed as we extend a programming language with more features.

## 9.2 A Minimalistic Imperative Language

WHILE<sup>1</sup> is a minimalistic imperative language with the following grammar:

```
S::= skip (no-op)
| x:= a (assignment)
| S; S (sequential composition)
| if b then Selse S (conditional statement)
| while b do S (while loop)
```

where S stands for a statement (also called command or program), x for a program variable, a for an arithmetic expression, and b for a Boolean expression.

In our grammar, we deliberately leave the syntax of arithmetic and Boolean expressions unspecified. In Lean, we have the choice:

- We can use a type such as AExp from Section 2.1 and similarly for Booleans.
- We can simply decide that an arithmetic expression is a function from states to numbers (e.g., State  $\to \mathbb{N}$ ) and a Boolean expression is a predicate over states (e.g., State  $\to$  Bool or State  $\to$  Prop). A *State* is a mapping from program variables to values. Thus, x + y + 1 would be represented by the function fun s : State  $\mapsto$  s "x" + s "y" + 1, and a  $\neq$  b would be represented by the predicate fun s : State  $\mapsto$  s "a"  $\neq$  s "b".

These two options correspond to the difference between deep and shallow embeddings. A *deep embedding* of some syntax (expression, formula, program, etc.) consists of an abstract syntax tree specified in the proof assistant (e.g., AExp) with a semantics (e.g., eval). In contrast, a *shallow embedding* simply reuses the corresponding mechanisms from the logic (e.g., functions and predicates).

A deep embedding allows us to reason about a program's syntax. A shallow embedding is more lightweight, because we can use it directly, without having to define a semantics. A shallow embedding is its own semantics.

In Chapter 7, we used a shallow embedding of effectful programs. Here, we will use a deep embedding of programs (which we find interesting and want to study closely) and a shallow embedding of arithmetic and Boolean expressions (which we find less interesting). Our Lean definition of programs follows:

<sup>&</sup>lt;sup>1</sup>Fans of backronyms might enjoy this one: Weak Hypothetical Imperative Language Example.

The infix syntax S; T abbreviates Stmt.seq S T.

The correspondence between the inductive type's constructors and the WHILE grammar rules should be clear. Variables are represented by strings. The type State is defined as  $String \to \mathbb{N}$ , a mapping from variable names to values. For simplicity, our program variables are all of type natural number, and all possible variable names exist in the state and are assigned a value.

The following small program illustrates deep embedding:

```
def sillyLoop : Stmt :=
  Stmt.whileDo (fun s → s "x" > s "y")
  (Stmt.skip;
  Stmt.assign "x" (fun s → s "x" - 1))
```

## 9.3 Big-Step Semantics

An *operational semantics* corresponds to an idealized interpreter. There are two main variants: big-step semantics and small-step semantics. We will start by giving a big-step semantics to our WHILE language.

In a big-step operational semantics (also called natural semantics), judgments have the form  $(S, s) \Rightarrow t$  and the following intuitive interpretation:

Starting in a state s, executing S may terminate in the state t.

For deterministic languages, since programs always have a single outcome, "may terminate" means the same as "must terminate" or "terminates."

In accordance with the definition of WHILE programs, a state s is a function of type  $String \to \mathbb{N}$ . An example judgment follows:

$$(x := x + y; y := 0, [x \mapsto 3, y \mapsto 5]) \Longrightarrow [x \mapsto 8, y \mapsto 0]$$

We use the informal notation  $[x \mapsto 3, y \mapsto 5]$  to represent the function fun  $v \mapsto$  if v = "x" then 3 else if v = "y" then 5 else 0 and similarly for  $[x \mapsto 8, y \mapsto 0]$ . Intuitively, the judgment holds.

The traditional way to specify such a semantics is through a formal system of derivation rules, in the style of the typing rules presented in Sections 1.3 and 4.6. The derivation rules for big-step semantics judgments are given below. The rules can be seen as an idealized interpreter for WHILE programs.

In the rules, a s denotes the value of arithmetic expression a in state s, and similarly for b s. Moreover, the syntax  $s[x \mapsto n]$  represents the state that is identical to s except that it maps the variable x to n. Formally:

$$s[x \mapsto n] = (fun \ v \mapsto if \ v = x \ then \ n \ else \ s \ v)$$

This syntax is provided by LoVelib.

As an exercise, let us derive the example judgment above. Let  $s := [x \mapsto 3, y \mapsto 5]$ ,  $t := [x \mapsto 8, y \mapsto 5]$ , and  $u := [x \mapsto 8, y \mapsto 0]$ . Then we have

$$\frac{\overline{(x := x + y, s) \Rightarrow t} \xrightarrow{ASSIGN} \overline{(y := 0, t) \Rightarrow u} \xrightarrow{ASSIGN} SEQ}{(x := x + y; y := 0, s) \Rightarrow u}$$

The derivation rules can be read intuitively. Consider SEQ:

If (1) executing S in state s leads to state t and (2) executing T in state t leads to state u, then executing the sequential composition S; T in state s leads to state u.

The conditions (1) and (2) correspond to the two premises of SEQ.

The most complicated rule is undoubtedly WHILE-TRUE. Intuitively, it can be understood as follows:

Assume condition b is true in state s. If (1) executing S in state s leads to state t and (2) executing while b do S from state t leads to state u, then executing while b do S in state s leads to state u.

Another way to think about WHILE-TRUE is in terms of loop unrolling. If the loop condition is true, while b do S is equivalent to the compound statement S; while b do S. The two premises of WHILE-TRUE correspond to the two premises of the instance of the SEQ rule for S; while b do S.

In Lean, a big-step semantics judgment is represented by an inductive predicate whose introduction rules closely follow the derivation rules above:

```
inductive BigStep : Stmt × State → State → Prop where
| skip (s) :
    BigStep (Stmt.skip, s) s
| assign (x a s) :
    BigStep (Stmt.assign x a, s) (s[x → a s])
| seq (S T s t u) (hS : BigStep (S, s) t)
        (hT : BigStep (T, t) u) :
    BigStep (S; T, s) u
| if_true (B S T s t) (hcond : B s)
        (hbody : BigStep (S, s) t) :
    BigStep (Stmt.ifThenElse B S T, s) t
```

```
| if_false (B S T s t) (hcond : ¬ B s)

    (hbody : BigStep (T, s) t) :

BigStep (Stmt.ifThenElse B S T, s) t

| while_true (B S s t u) (hcond : B s)

    (hbody : BigStep (S, s) t)

    (hrest : BigStep (Stmt.whileDo B S, t) u) :

BigStep (Stmt.whileDo B S, s) u

| while_false (B S s) (hcond : ¬ B s) :

BigStep (Stmt.whileDo B S, s) s
```

We use implicit arguments, within curly braces, for many of the variables corresponding to the derivation rule's mathematical variables.

Using an inductive predicate as opposed to a recursive function allows us to cope with nontermination (a diverging while) and, for languages richer than WHILE, nondeterminism. It also arguably provides a nicer syntax, closer to the judgment rules that are traditionally used in the scientific literature. If we were instead to attempt a recursive definition such as

we would face nontermination of the Stmt.whileDo case. Indeed, since the program Stmt.whileDo (fun  $\_ \mapsto \mathsf{True}$ ) Stmt.skip loops forever, trying to evaluate it using eval would never return.

Equipped with a big-step semantics, we can reason about concrete programs such as the one defined in Section 9.2 and prove theorems such as the following:

```
theorem sillyLoop_from_1_BigStep :
  (sillyLoop, (fun _ → o)["x" → 1]) ⇒ (fun _ → o) :=
  by
    rw [sillyLoop]
    apply BigStep.while_true
    { aesop }
    { apply BigStep.seq
        { apply BigStep.skip }
        { apply BigStep.assign } }
        { simp
            apply BigStep.while_false
            simp }
```

# 9.4 Properties of the Big-Step Semantics

Equipped with a big-step semantics, we can reason about concrete programs, proving theorems relating final states with initial states. Perhaps more interest-

ingly, we can prove properties of the programming language, such as determinism and nontermination.

We start with determinism. It may seem like a trivial property, but it is easy to mistype a rule and obtain nondeterminism. For example, in the rule for assignment, if we mistakenly write BigStep (Stmt.assign x a, s) ( $s[y \mapsto a s]$ ) with y instead of x, we can suddenly use the rule to modify any variable y we please. In other words, an execution of the program could modify any variable at random. So let us verify that our WHILE language really is deterministic:

```
theorem BigStep_deterministic {Ss l r} (hl : Ss \Longrightarrow l) (hr : Ss \Longrightarrow r) : l = r
```

The Lean proof is in the demonstration file associated with this chapter. For technical reasons, the pair (S, s) is represented by a single variable Ss. We content ourselves with an informal proof sketch:

The proof is by rule induction over  $(S, s) \Rightarrow l$ .

```
CASE SKIP: To have (skip, s) \Longrightarrow l or (skip, s) \Longrightarrow r, we need l = r = s.
```

CASE ASSIGN: Similar to SKIP.

CASE SEQ: We have the hypotheses (S, s)  $\Longrightarrow$  t, (T, t)  $\Longrightarrow$  l, (S, s)  $\Longrightarrow$  t', and (T, t')  $\Longrightarrow$  r and the induction hypotheses  $\forall$ r, (S, s)  $\Longrightarrow$  r  $\rightarrow$  t = r and  $\forall$ r, (T, t)  $\Longrightarrow$  r  $\rightarrow$  l = r. From the first induction hypothesis together with (S, s)  $\Longrightarrow$  t', we derive t = t'. From the second induction hypothesis together with (T, t')  $\Longrightarrow$  r, we derive l = r.

CASE IF-TRUE: Since b s is true, (if b then S else T, s)  $\Longrightarrow$  r can only have been derived using IF-TRUE and thus (S, s)  $\Longrightarrow$  r. The induction hypothesis is  $\forall$ r, (S, s)  $\Longrightarrow$  r  $\rightarrow$  l = r. We can apply (S, s)  $\Longrightarrow$  r to it to obtain l = r.

CASE IF-FALSE: Similar to IF-TRUE.

CASE WHILE-TRUE: Similar to SEQ.

CASE WHILE-FALSE: Similar to SKIP.

Given that the WHILE language is deterministic, for the big-step semantics, termination would amount to the following:

```
theorem BigStep_terminates \{S \ s\}:
\exists t, (S, s) \Rightarrow t
```

This property means that for every statement S and state s, there exists a state t such that executing S starting in s may terminate in t. Because WHILE is deterministic, "may terminate" means the same as "must terminate." However, the property does not hold.

When reasoning about an inductive predicate, it is often convenient to use inversion rules (Section 6.5). Accordingly, we prove the following rules:

```
@[simp] theorem BigStep_skip_Iff {s t} : (Stmt.skip, s) \implies t \leftrightarrow t = s
```

```
@[simp] theorem BigStep_assign_Iff {x a s t} :
  (Stmt.assign x a, s) \Rightarrow t \leftrightarrow t = s[x \mapsto a s]
a[simp] theorem BigStep_seq_Iff {S T s u} :
  (S; T, s) \Rightarrow u \leftrightarrow (\exists t, (S, s) \Rightarrow t \land (T, t) \Rightarrow u)
a[simp] theorem BigStep_if_Iff {B S T s t} :
  (Stmt.ifThenElse B S T, s) \Longrightarrow t \leftrightarrow
  (B s \land (S, s) \Rightarrow t) \lor (\neg B s \land (T, s) \Rightarrow t)
theorem BigStep_while_Iff {B S s u} :
  (Stmt.whileDo B S, s) \Longrightarrow u \leftrightarrow
  (\exists t, B s \land (S, s) \Rightarrow t \land (Stmt.whileDo B S, t) \Rightarrow u)
  \vee (\neg B s \wedge u = s)
a[simp] theorem BigStep_while_true_Iff {B S s u}
     (hcond : B s) :
  (Stmt.whileDo B S, s) \Longrightarrow u \leftrightarrow
  (\exists t, (S, s) \Rightarrow t \land (Stmt.whileDo B S, t) \Rightarrow u)
a[simp] theorem BigStep_while_false_Iff {B S s t}
     (hcond : \neg B s) :
  (Stmt.whileDo B S, s) \Longrightarrow t \leftrightarrow t = s
```

We add the rules to the simp set except for BigStep\_while\_Iff because it would make simp loop.

# 9.5 Small-Step Semantics

A limitation of big-step semantics is that they do not let us reason about intermediate states. From a judgment  $(s, s) \Longrightarrow t$ , all we see is the initial state s and the final state t. This is too coarse-grained to reason about multithreaded programs, where several processes can interact with each other's intermediate states. Moreover, for nondeterministic languages, big-step semantics offer no general way to express termination: A judgment indicates a possibility (executing s in state s may result in state t), not a necessity.

Small-step operational semantics provide a finer view. The transition predicate  $\Rightarrow$  has type Stmt  $\times$  State  $\rightarrow$  Stmt  $\times$  State  $\rightarrow$  Prop. Intuitively, (S, s)  $\Rightarrow$  (T, t) means that executing one step of program S in state s leaves the program T to be executed, in state t. If there is nothing left to be executed, we put skip.

An execution is a finite or infinite chain  $(S_0, s_0) \Rightarrow (S_1, s_1) \Rightarrow \cdots$  of "small"  $\Rightarrow$  steps. A pair (S, s) is called a *configuration*; it is *final* if no transition of the form  $(S, s) \Rightarrow (T, t)$  is possible, for any (T, t). A possible execution follows:

```
(x := x + y; y := 0, [x \mapsto 3, y \mapsto 5])
\Rightarrow (skip; y := 0, [x \mapsto 8, y \mapsto 5])
\Rightarrow (y := 0, [x \mapsto 8, y \mapsto 5])
\Rightarrow (skip, [x \mapsto 8, y \mapsto 0])
```

If we take the analogy of a computer processor, the S component of a configuration (S, s) can be thought of as a program counter, which indicates which

instructions should be executed next. The step-by-step execution of a program resembles running a program in a debugger with a breakpoint at each step.

The valid small-step judgments are given by derivation rules:

```
\frac{(\text{S}, \text{s}) \Rightarrow (\text{skip}, \text{s}[\text{x} \mapsto \text{a} \text{s}])}{(\text{S}, \text{S}) \Rightarrow (\text{S}', \text{s}')} \text{SEQ-STEP}
\frac{(\text{S}, \text{T}, \text{s}) \Rightarrow (\text{S}', \text{T}, \text{s}')}{(\text{skip}; \text{T}, \text{s}) \Rightarrow (\text{T}, \text{s})} \text{SEQ-SKIP}
\frac{(\text{skip}; \text{T}, \text{s}) \Rightarrow (\text{T}, \text{s})}{(\text{if b then S else T}, \text{s}) \Rightarrow (\text{S}, \text{s})} \text{IF-TRUE if b s is true}
\frac{(\text{if b then S else T}, \text{s}) \Rightarrow (\text{S}, \text{s})}{(\text{if b then S else T}, \text{s}) \Rightarrow (\text{T}, \text{s})} \text{IF-FALSE if b s is false}
\frac{(\text{while b do S}, \text{s}) \Rightarrow (\text{if b then (S; while b do S) else skip, s})}{(\text{while b do S}, \text{s}) \Rightarrow (\text{if b then (S; while b do S) else skip, s})} \text{WHILE}
```

The rules are reminiscent of the tennis game transition system of Section 6.1.2. These, too, specified small steps: from 0-0 to 15-0, to 15-15, and so on.

Unlike in the big-step semantics, there is no rule for skip in the small-step semantics. This is because a configuration of the form (skip, s) is considered final; skip is understood as the statement whose execution is trivial. By inspection of the rules, we can convince ourselves that a configuration is final if and only if its first component is skip.

Two rules concern sequential composition S; T. The first rule is applicable if some progress can be made executing S. But if S is skip, no progress can be made and the second rule applies.

The rules for if check the condition b and, depending on its truth value, put the then or the else branch as the remaining computation to perform.

For the while loop, there is a single unconditional rule that expands one iteration of the loop, introducing an if statement. It is then the role of the IF-TRUE and IF-FALSE rules to process the if. In the IF-TRUE case, we eventually reach the while loop again. This can continue forever for infinite loops.

In Lean, the small-step semantics is specified as follows:

```
inductive SmallStep : Stmt × State → Stmt × State → Prop
  where
| assign (x a s) :
    SmallStep (Stmt.assign x a, s) (Stmt.skip, s[x → a s])
| seq_step (S S' T s s') (hS : SmallStep (S, s) (S', s')) :
    SmallStep (S; T, s) (S'; T, s')
| seq_skip (T s) :
    SmallStep (Stmt.skip; T, s) (T, s)
| if_true (B S T s) (hcond : B s) :
    SmallStep (Stmt.ifThenElse B S T, s) (S, s)
| if_false (B S T s) (hcond : ¬ B s) :
    SmallStep (Stmt.ifThenElse B S T, s) (T, s)
| whileDo (B S s) :
```

```
SmallStep (Stmt.whileDo B S, s)
  (Stmt.ifThenElse B (S; Stmt.whileDo B S) Stmt.skip, s)
```

Based on a small-step semantics, we can *define* a big-step semantics as follows:

```
(S, s) \Rightarrow t if and only if (S, s) \Rightarrow * (skip, t)
```

where p\* denotes the reflexive transitive closure (RTC) of a binary predicate p. Alternatively, if we have already defined a big-step semantics, we can *prove* the above equivalence theorem to *validate* our definitions.

The main disadvantage of small-step semantics is that we now have two predicates,  $\Rightarrow$  and  $\Rightarrow$ \*, and the derivation rules and proofs tend to be more complicated than with big steps. This is clearly visible on the following example, in which we need to apply the theorem

```
RTC.head: ?R?a?b \rightarrow RTC?R?b?c \rightarrow RTC?R?a?c
```

once for each small step:

```
theorem sillyLoop_from_1_SmallStep :
  (sillyLoop, (fun \rightarrow 0)["x" \rightarrow 1]) \Rightarrow *
  (Stmt.skip, (fun \rightarrow 0)) :=
    rw [sillyLoop]
    apply RTC.head
    { apply SmallStep.whileDo }
    { apply RTC.head
      { apply SmallStep.if_true
        aesop }
      { apply RTC.head
        { apply SmallStep.seq_step
          apply SmallStep.seq_skip }
        { apply RTC.head
          { apply SmallStep.seq_step
            apply SmallStep.assign }
          { apply RTC.head
             { apply SmallStep.seq_skip }
             { apply RTC.head
               { apply SmallStep.whileDo }
               { apply RTC.head
                 { apply SmallStep.if_false
                   simp }
                 { simp
                   apply RTC.refl } } } } }
```

# 9.6 Properties of the Small-Step Semantics

We can prove that a configuration (S, s) is final if and only if S = skip. Doing so ensures that we have not forgotten a derivation rule and hence that the small-steps semantics cannot get stuck. The theorem statement is as follows:

```
theorem SmallStep_final (S s): (\neg \exists T \ t, (S, s) \Rightarrow (T, t)) \leftrightarrow S = Stmt.skip
```

The proof is by structural induction on S.

Like the big-step semantics, the small-step semantics is deterministic:

```
theorem SmallStep_deterministic {Ss Ll Rr} (hl : Ss \Rightarrow Ll) (hr : Ss \Rightarrow Rr) : Ll = Rr
```

The proof is by rule induction on hl or hr.

For the small-step semantics, a configuration  $(S_0, s_0)$  terminates if all executions starting in it are finite:  $(S_0, s_0) \Rightarrow (S_1, s_1) \Rightarrow \cdots \Rightarrow (S_n, s_n)$ . It is nonterminating if there exists an infinite chain  $(S_0, s_0) \Rightarrow (S_1, s_1) \Rightarrow \cdots$ . The programming language as a whole is terminating if and only if all its configurations terminate. It is easy to show that the WHILE language is nonterminating, by taking  $S_0 := Stmt.whileDo(fun_ \mapsto True)Stmt.skip.$  For any  $s_0$ , we then have

```
(S_0, S_0) \Rightarrow (S_0, S_0) \Rightarrow (S_0, S_0) \Rightarrow \cdots
```

We can define inversion rules about the small-step semantics, such as these:

```
theorem SmallStep_skip {S s t}:
\neg ((Stmt.skip, s) \Rightarrow (S, t))
@[simp] theorem SmallStep_seq_Iff {S T s Ut}:
(S; T, s) \Rightarrow Ut \leftrightarrow
(\exists S' t, (S, s) \Rightarrow (S', t) \land Ut = (S'; T, t))
\lor (S = Stmt.skip \land Ut = (T, s))
@[simp] theorem SmallStep_if_Iff {B S T s Us}:
(Stmt.ifThenElse B S T, s) \Rightarrow Us \leftrightarrow
(B s \land Us = (S, s)) \lor (\neg B s \land Us = (T, s))
```

A more fundamental result is the equivalence between the big-step and the small-step semantics:

```
theorem BigStep_Iff_RTC_SmallStep {Ss t} : Ss \implies t \leftrightarrow Ss \implies * (Stmt.skip, t)
```

Recall that  $\Rightarrow *$  denotes the reflexive transitive closure of the small-step predicate  $\Rightarrow$ . The theorem's proof is beyond the scope of this course. We refer to Chapter 7 of *Concrete Semantics: With Isabelle/HOL* [23] or to the demonstration file for this chapter.

# **Chapter 10**

# **Hoare Logic**

If operational semantics corresponds to an idealized interpreter, *Hoare logic* corresponds to an idealized verifier. Hoare logic can be used to specify the semantics of a programming language, but it is particularly convenient to reason about concrete programs and prove them correct. It is named after its inventor, Charles Antony Richard (Tony) Hoare. Hoare logic is also called *axiomatic semantics*.

This chapter is heavily inspired by Chapter 12 of Concrete Semantics: With Isabelle/HOL [23].

# 10.1 Hoare Triples

Hoare logic is a framework for deducing valid correctness formulas in a mechanical way, using a set of derivation rules. It allows us to reason directly about a program's syntax, without concerning ourselves with its operational semantics. The approach is mechanical in the sense that the applicability of a derivation rule can easily be checked.

We start by introducing Hoare logic abstractly, without any connection to Lean. In a second step, we will see how we can embed Hoare logic judgments in Lean. The basic judgments of Hoare logic are called *Hoare triples*. They have the form {P} S {Q}, where S is a WHILE statement, and P and Q are logical formulas over the program variables. For the moment, we imagine the formulas as syntactic objects built using the familiar connectives and quantifiers. The intended meaning of a Hoare triple is as follows:

If the precondition P is true before S is executed and the execution terminates normally, the postcondition Q is true at termination.

This is a partial correctness statement: The program is correct if it terminates normally; otherwise, the program might behave arbitrarily. For WHILE programs, the only way not to terminate normally is to enter an infinite loop. For other programming languages, infinite recursion and run-time errors such as division by zero may also result in divergence or abnormal termination.

Intuitively, all of the Hoare triples below should be valid:

{True} 
$$b := 4 \{b = 4\}$$
  
{ $a = 2$ }  $b := 2 * a \{a = 2 \land b = 4\}$   
{ $b \ge 5$ }  $b := b + 1 \{b \ge 6\}$ 

{False} skip {b = 10}  
{True} while 
$$i \neq 10$$
 do  $i := i + 1$  { $i = 10$ }

The first three Hoare triples should be fairly natural. The fourth triple is vacuously true, since the precondition False can never be met. The part "If the precondition P is true" of the definition of Hoare triple is always false; hence the triple is true. The triple is equivalent to the proposition False  $\rightarrow$  b = 10, which holds for any value of b. As for the fifth triple, there are no guarantees that control will escape the loop (if i > 10 initially), but if it does escape, then the loop's condition must be false and hence we have the postcondition i = 10.

The following triples are bizarre but interesting:

The first two triples are true of any statement S (and therefore pointless): The precondition is never satisfied, so any postcondition holds vacuously. The third triple is always true as well, regardless of S. The fourth triple is true if S never terminates (e.g., S := while True do skip); otherwise, it is false.

#### 10.2 Hoare Rules

Below we give a complete set of derivation rules for reasoning about WHILE programs:

In the Assign rule, the expression Q[a/x] denotes the condition Q in which all occurrences of x are replaced by a. The rule is sometimes presented as

$$\frac{}{\{Q[a]\} \times := a \{Q[x]\}} ASSIGN$$

where [x] factors out the occurrences of x in Q.

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The Assign rule may seem counterintuitive because it works backwards: From the postcondition, it computes a precondition. Nevertheless, it correctly captures the semantics of the assignment statement, as illustrated below:

$$\{0 = 0\} \ x := 0 \ \{x = 0\}$$
  
 $\{0 = 0 \land y = 5\} \ x := 0 \ \{x = 0 \land y = 5\}$   
 $\{x + 1 \ge 5\} \ x := x + 1 \ \{x \ge 5\}$ 

Using elementary arithmetic, we can simplify the computed preconditions; for example, o = o is equivalent to True, and  $x + 1 \ge 5$  is equivalent to  $x \ge 4$ .

The SEQ rule requires us to come up with an intermediate condition R that holds after executing S and before executing T. Here is an example of SEQ in action:

$$\frac{ \{a = 2\} \ b := a \ \{b = 2\} }{ \{a = 2\} \ b := a; \ c := b \ \{c = 2\} }$$
 ASSIGN 
$$\frac{ \{b = 2\} \ c := b \ \{c = 2\} }{ SEQ }$$

The WHILE rule is the most intricate. The condition P is called an *invariant*. It is the pre- and the postcondition of the loop itself but also of its body. The body's precondition is strengthened with the knowledge that B must be true before executing the body. Similarly, the loop's postcondition is strengthened with the knowledge that B must be false when the loop exits.

Consider an execution of a loop with n iterations. Suppose that the initial state is  $s_0$  and that the state after the ith iteration of the loop is  $s_i$ . Then the following conditions will hold:

$$P s_0 \qquad P s_1 \wedge B s_1 \qquad \cdots \qquad P s_{n-1} \wedge B s_{n-1} \qquad P s_n \wedge \neg B s_n$$

If n = 0, we immediately have that  $P s_0 \land \neg B s_0$  and never enter the loop.

Conseq is the only rule that has logical formulas among its premises, as opposed to Hoare triples. These conditions must be discharged, whether using pen and paper or a proof assistant. Conseq can be used to strengthen a precondition (i.e., make it more restrictive), weaken a postcondition (i.e., make it less restrictive), or both. An example derivation follows:

$$\frac{x > 8 \to x > 4}{\{x > 4\} \ y := x \ \{y > 4\}} \xrightarrow{ASSIGN} y > 4 \to y > 0}{\{x > 8\} \ y := x \ \{y > 0\}} CONSEQ$$

Reading the tree from top to bottom, we have strengthened the triple's precondition from x > 4 to x > 8 and weakened the postcondition from y > 4 to y > 0. We can also read the tree from the bottom up: To prove the triple  $\{x > 8\}$  y := x  $\{y > 0\}$ , it suffices to prove  $\{x > 4\}$  y := x  $\{y > 4\}$ , where the precondition is weakened and the postcondition is strengthened.

Except for CONSEQ, the rules are *syntax-driven*: We know which rule to apply in each case, simply by inspecting the statement at hand. For an assignment, we apply ASSIGN; for a while loop, we apply WHILE; and so on.

The rules SEQ, IF, and CONSEQ are bidirectional: Their conclusions are of the form  $\{P\}$  ...  $\{Q\}$  for distinct mathematical variables P, Q. This can make them convenient to apply. By combining the other rules with CONSEQ, we can derive bidirectional variants:

$$\frac{P \rightarrow Q}{\{P\} \text{ skip } \{Q\}} \text{SKIP'}$$

$$\frac{P \rightarrow Q[a/x]}{\{P\} \text{ } x := a \text{ } \{Q\}} \text{ASSIGN'}$$

$$\frac{\{P \land B\} \text{ } S \text{ } \{P\} \text{ } P \land \neg B \rightarrow Q}{\{P\} \text{ while } B \text{ do } S \text{ } \{Q\}} \text{WHILE'}$$

As an exercise, you could try to derive each of these rules from SKIP, ASSIGN, or WHILE in conjunction with CONSEQ.

# 10.3 A Semantic Approach to Hoare Logic

A natural way to encode Hoare logic in Lean would be to proceed as we have done for the big- and small-step semantics: Define a syntactic notion of Hoare triple and an inductive predicate, with one introduction rule for each core Hoare rule; to represent pre- and postconditions, use predicates on states ( $State \rightarrow Prop$ ). Then we could prove soundness with respect to the big-step semantics, meaning the following: Whenever {P} S {Q} is derivable, if P s and (S, s)  $\Longrightarrow$  t, then Q t. This is merely a logical rendition of the intuitive meaning of a Hoare triple:

If the precondition P is true before S is executed (i.e., P s) and the execution terminates normally (i.e.,  $(S, s) \Longrightarrow t$ ), the postcondition Q is true at termination (i.e., Q t).

Instead of pursuing this approach, we propose to define Hoare triples semantically in Lean, in terms of the big-step semantics, so that they are *correct by definition*. Then we will derive the Hoare rules as theorems, instead of stating them as introduction rules. In conjunction with the use of predicates to represent formulas, this approach is resolutely semantic.

Hoare triples (for partial correctness) are defined as follows:

```
def PartialHoare (P : State \rightarrow Prop) (S : Stmt)
(Q : State \rightarrow Prop) : Prop :=
\foralls t, P s \rightarrow (S, s) \Longrightarrow t \rightarrow Q t
```

Instead of writing PartialHoare P S Q, we introduce some syntactic sugar to allow  $\{* P *\} (S) \{* Q *\}$ , which is closer to the informal syntax  $\{P\} S \{Q\}$ .

The core Hoare rules are stated as follows:

```
theorem skip_intro {P} :
    {* P *} (Stmt.skip) {* P *}

theorem assign_intro (P) {x a} :
    {* fun s → P (s[x → a s]) *} (Stmt.assign x a) {* P *}

theorem seq_intro {P Q R S T} (hS : {* P *} (S) {* Q *})
    (hT : {* Q *} (T) {* R *}) :
    {* P *} (S; T) {* R *}
```

```
theorem if_intro {B P Q S T}
    (hS : {* fun s → P s ∧ B s *} (S) {* Q *})
    (hT : {* fun s → P s ∧ ¬ B s *} (T) {* Q *}) :
    {* P *} (Stmt.ifThenElse B S T) {* Q *}

theorem while_intro (P) {B S}
    (h : {* fun s → P s ∧ B s *} (S) {* P *}) :
    {* P *} (Stmt.whileDo B S) {* fun s → P s ∧ ¬ B s *}

theorem consequence {P P' Q Q' S}
    (h : {* P *} (S) {* Q *}) (hp : ∀s, P' s → P s)
    (hq : ∀s, Q s → Q' s) :
    {* P' *} (S) {* Q' *}
```

All of the above theorems have proofs based on the big-step semantics. Some of the triples—for example, the precondition in <code>assign\_intro</code>—need to refer to the state. We then use an anonymous function to access it. Recall that P and fun  $s\mapsto P$  s are equal (by  $\eta$ -conversion). Moreover, for a premise written informally as  $P\to Q$ , in Lean we must write  $\forall s$ , P s  $\to Q$  s. As in Chapter 9, the syntax  $s[x\mapsto n]$  in the assignment rule denotes the state that is identical to s except that it maps x to n.

The following convenience rules can be derived from the core rules:

```
theorem consequence left (P') {P Q S}
    (h : \{*\ P\ *\}\ (S)\ \{*\ Q\ *\}) (hp : \forall s,\ P'\ s \rightarrow P\ s) :
  {* P' *} (S) {* Q *}
theorem consequence_right (Q) {Q' P S}
    (h : \{* P *\} (S) \{* Q *\}) (hq : \forall s, Q s \rightarrow Q' s) :
  {* P *} (S) {* Q' *}
theorem skip_intro' \{P Q\} (h : \forall s, P s \rightarrow Q s) :
  {* P *} (Stmt.skip) {* Q *}
theorem assign_intro' {P Q x a}
    (h : \foralls, P s \rightarrow Q (s[x \mapsto a s])):
  {* P *} (Stmt.assign x a) {* Q *}
theorem seq_intro' {P Q R S T} (hT : {* Q *} (T) {* R *})
    (hS : {* P *} (S) {* Q *}) :
  {* P *} (S; T) {* R *}
theorem while_intro' {B P Q S} (I)
    (hS : \{* fun s \mapsto I s \land B s *\} (S) \{* I *\})
    (hP: \foralls, Ps \rightarrow Is)
    (hQ : \foralls, \neg B s \rightarrow I s \rightarrow Q s) :
  {* P *} (Stmt.whileDo B S) {* Q *}
```

Using the bidirectional assign\_intro', we can derive a forward version of the assignment rule:

```
theorem assign_intro_forward (P) {x a} : 
 {* P *} 
 (Stmt.assign x a) 
 {* fun s \mapsto \exists n_0, P (s[x \mapsto n_0]) \land s x = a (s[x \mapsto n_0]) *} := 
 by 
 apply assign_intro' 
 intro s hP 
 apply Exists.intro (s x) 
 simp [*]
```

The variable  $n_0$  stands for the value of x before the assignment. Hence, in the postcondition,  $s[x \mapsto n_0]$  denotes the state before the assignment. Since P is a precondition, we have  $P(s[x \mapsto n_0])$ . In addition, the new value of x, given by s x, must be equal to the value of the expression a evaluated in the old state  $s[x \mapsto n_0]$ .

The forward rule is less convenient than the backward rule, because the postcondition contains an existential quantifier. It is possible to state a backward rule in a similar style, revealing a hidden symmetry:

```
theorem assign_intro_backward (Q) \{x \ a\}: \{* \ fun \ s \mapsto \exists n', \ Q \ (s[x \mapsto n']) \land n' = a \ s \ *\} (Stmt.assign x \ a) \{* \ Q \ *\}
```

Notice that this existential quantifier can be eliminated using a one-point rule (Section 4.3). We then get the familiar backward rule assign\_intro, with fun  $s \mapsto P(s[x \mapsto a \ s])$  as the precondition.

# 10.4 First Program: Exchanging Two Variables

Let us employ Hoare logic to verify our first program: a three-line program that exchanges the values of its variables a and b, using t for temporary storage. The program is defined as follows:

```
def SWAP : Stmt :=
  Stmt.assign "t" (fun s \mapsto s "a");
  Stmt.assign "a" (fun s \mapsto s "b");
  Stmt.assign "b" (fun s \mapsto s "t")
```

The correctness statement is as follows:

```
theorem SWAP_correct (a_0 \ b_0 : \mathbb{N}) : \{* \ fun \ s \mapsto s \ "a" = a_0 \land s \ "b" = b_0 *\}  (SWAP) \{* \ fun \ s \mapsto s \ "a" = b_0 \land s \ "b" = a_0 *\}
```

The logical variables  $a_o$  and  $b_o$  "freeze" the initial value of the program variables a and b so that we can refer to them in the postcondition. After all, it would make no sense to use fun  $s\mapsto s$  "a" = s "b"  $\wedge$  s "b" = s "a" as the postcondition.

The correctness proof follows:

```
by
   apply PartialHoare.seq_intro'
```

```
apply PartialHoare.seq_intro'
apply PartialHoare.assign_intro
apply PartialHoare.assign_intro
apply PartialHoare.assign_intro'
aesop
```

The applications of the sequential composition and assignment rules are guided by the program's syntax. There are two sequential compositions and three assignments in the program and therefore as many invocations of the corresponding rules. We end up with a very aesthetically challenged subgoal:

```
\begin{array}{l} \mbox{$\vdash$ } \forall s: State, \\ s \mbox{$"a" = a_0 \land s $"b" = b_0 \to$} \\ s[\mbox{$"t" \mapsto s $"a"]["a" \mapsto s[\mbox{$"t" \mapsto s $"a"] "b"]$} \\ [\mbox{$"b" \mapsto s[\mbox{$"t" \mapsto s $"a"]["a" \mapsto s[\mbox{$"t" \mapsto s $"a"] "b"]$}} \\ s[\mbox{$"t" \mapsto s $"a"]["a" \mapsto s[\mbox{$"t" \mapsto s $"a"] "b"]$} \\ [\mbox{$"b" \mapsto s[\mbox{$"t" \mapsto s $"a"]["a" \mapsto s[\mbox{$"t" \mapsto s $"a"] "b"] "t"] "b" = a_0$} \end{array}
```

Fortunately, simp [\*] at \* can reduce the subgoal dramatically, and aesop can even prove it fully automatically.

# 10.5 Second Program: Adding Two Numbers

Our second example computes m + n, leaving the result in m, using only these primitive operations: k + 1, k - 1, and  $k \neq 0$  (for arbitrary k):

```
def ADD : Stmt :=
  Stmt.whileDo (fun s \mapsto s "n" \neq 0)
  (Stmt.assign "n" (fun s \mapsto s "n" - 1);
  Stmt.assign "m" (fun s \mapsto s "m" + 1))
```

Because of the presence of a while loop, the proof is more involved:

```
theorem ADD_correct (n_0 \ m_0 : \mathbb{N}) : {* fun s \mapsto s "n" = n_0 \land s "m" = m_0 *} (ADD) {* fun s \mapsto s "n" = 0 \land s "m" = n_0 + m_0 *} := PartialHoare.while_intro' (fun s \mapsto s "n" + s "m" = n_0 + m_0) (by apply PartialHoare.seq_intro' { apply PartialHoare.assign_intro } { apply PartialHoare.assign_intro' aesop }) (by aesop)
```

The first step is to apply the derived while rule with a loop invariant. Our invariant is that the sum of the program variables n and m must be equal to the desired mathematical result  $n_0 + m_0$ , where  $n_0$  and  $m_0$  correspond to the initial values of n and m, as required by the precondition.

How did we come up with this invariant? Even for a simple loop, finding a suitable invariant can be challenging. The difficulty is that the invariant must

- 1. be true before we enter the loop;
- 2. remain true after each iteration of the loop if it was true before the iteration;
- 3. be strong enough to imply the desired loop postcondition.

An invariant such as True meets requirements 1 and 2 but usually not 3. Similarly, False meets requirements 2 and 3 but not 1. In practice, invariants tend to be of the form

work done + work remaining = desired result

where + stands for some appropriate operator (not necessarily addition). When we enter the loop, work done will often be  $\circ$  (or some other appropriate "zero" value), and the invariant becomes

work remaining = desired result

This invariant would have to be provable at the beginning of the loop—either from the postcondition of the previous statement or from the desired precondition of the entire program if there is no previous statement. When we exit the loop, work remaining should be  $\odot$  (or some variant thereof), and the invariant becomes

work done = desired result

Often, work done takes the form of a variable in which we accumulate the result, whereas work remaining is similar to desired result but depends on program variables whose values change inside the loop and accounts for work done.

For the ADD program's loop, the work done is m, the work remaining is n, and the desired result is  $n_0 + m_0$ . When entering the loop, the invariant  $m + n = n_0 + m_0$  holds because then  $m = m_0$  and  $n = n_0$ . (Unusually, the work done is not o for this example, because we reuse the input m as our accumulator, as an optimization.) When exiting the loop, we have n = 0, so the invariant becomes  $m = n_0 + m_0$ . We can retrieve the result from m.

The while\_intro' theorem is used directly as a proof term. It gives rise to three subgoals. The proofs that the invariant is implied by the desired precondition and that it implies the desired postcondition are trivial: They consist of a call to aesop. The only nontrivial subgoal is the condition that executing the body maintains the invariant.

For this example, Hoare logic really helps. Reasoning directly about the operational semantics would be inconvenient, because we would need induction to reason about the while loop. With Hoare logic, this induction is performed once and for all in the proof of the while\_intro rule.

#### 10.6 A Verification Condition Generator

Verification condition generators (VCGs) are programs that apply Hoare logic rules, producing verification conditions that must be proved manually. We can think of them as mechanical civil servants that take care of the Hoare logic bureaucracy. As users, we must provide strong enough loop invariants as annotations in our programs. Hundreds of program verification tools are based on these principles.

VCGs typically work backwards from the postcondition, using *backward* rules—rules stated to have an arbitrary Q as their postcondition. This works well because the central rule of Hoare logic—the assignment rule—is backward.

We can use Lean's metaprogramming framework to define a simple VCG. First, we introduce a constant called Stmt.invWhileDo that carries a user-supplied invariant I in addition to the loop condition B and the body S:

```
def Stmt.invWhileDo (I B : State \rightarrow Prop) (S : Stmt) : Stmt :=
   Stmt.whileDo B S
```

We provide two Hoare rules for the construct: a backward rule and a bidirectional rule. Both are justified in terms of the bidirectional while\_intro' rule:

```
theorem invWhile_intro {B I Q S}
    (hS : {* fun s → I s ∧ B s *} (S) {* I *})
    (hQ : ∀s, ¬ B s → I s → Q s) :
    {* I *} (Stmt.invWhileDo I B S) {* Q *} :=
    while_intro' I hS (by aesop) hQ

theorem invWhile_intro' {B I P Q S}
    (hS : {* fun s → I s ∧ B s *} (S) {* I *})
    (hP : ∀s, P s → I s) (hQ : ∀s, ¬ B s → I s → Q s) :
    {* P *} (Stmt.invWhileDo I B S) {* Q *} :=
    while_intro' I hS hP hQ
```

The above rules simply use the invariant annotation as their invariant. When using the framework, we must be careful to annotate all while loops with suitable invariants. If we specify a wrong invariant, we will face unprovable subgoals, indicating that we must revise the invariant.

The code of the VCG is fairly concise:

```
def matchPartialHoare : Expr \rightarrow Option (Expr \times Expr \times Expr)
  | (Expr.app (Expr.app (Expr.app
       (Expr.const ''PartialHoare _) P) S) Q) =>
    Option.some (P, S, Q)
  | _ =>
    Option.none
partial def vcg : TacticM Unit :=
    let goals ← getUnsolvedGoals
    if goals.length != o then
      let target ← getMainTarget
      match matchPartialHoare target with
      | Option.none
                               => return
      | Option.some (P, S, Q) =>
        if Expr.isAppOfArity S ''Stmt.skip o then
          if Expr.isMVar P then
            applyConstant ''PartialHoare.skip_intro
            applyConstant ''PartialHoare.skip_intro'
        else if Expr.isAppOfArity S ''Stmt.assign 2 then
          if Expr.isMVar P then
            applyConstant ''PartialHoare.assign_intro
          else
```

```
applyConstant ''PartialHoare.assign_intro'
        else if Expr.isAppOfArity S ''Stmt.seq 2 then
          andThenOnSubgoals
            (applyConstant ''PartialHoare.seq_intro') vcg
        else if Expr.isAppOfArity S ''Stmt.ifThenElse 3 then
          andThenOnSubgoals
            (applyConstant ''PartialHoare.if_intro) vcg
        else if Expr.isAppOfArity S ''Stmt.invWhileDo 3 then
          if Expr.isMVar P then
            andThenOnSubgoals
              (applyConstant ''PartialHoare.invWhile_intro) vcg
          else
            andThenOnSubgoals
              (applyConstant ''PartialHoare.invWhile_intro')
              vcg
        else
          failure
elab "vcg" : tactic =>
 vcg
```

The VCG extracts the first goal's target and inspects it. If it is a Hoare triple, the VCG inspects its precondition P and statement S. If the precondition is a metavariable (e.g., ?P), the VCG applies a backward rule (via the applyConstant function we defined in Section 8.7) if one exists because this will instantiate the metavariable. Otherwise, a bidirectional rule is used, with an arbitrary variable as its precondition, which can be matched against the target's precondition. For while loops, we only consider programs that use Stmt.invWhileDo, because we cannot guess the invariant programmatically.

The VCG calls itself recursively on all newly emerging subgoals for the compound statements (via the andThenOnSubgoals function we defined in Section 8.7).

# 10.7 Second Program Revisited: Adding Two Numbers

Using the verification condition generator, we can revisit the correctness proof for the ADD program presented above:

```
theorem ADD_correct_vcg (n_0 \ m_0 : \mathbb{N}): {* fun s \mapsto s "n" = n_0 \land s "m" = m_0 *} (ADD) {* fun s \mapsto s "n" = 0 \land s "m" = n_0 + m_0 *} := show {* fun s \mapsto s "n" = n_0 \land s "m" = m_0 *} (Stmt.invWhileDo (fun s \mapsto s "n" + s "m" = n_0 + m_0) (fun s \mapsto s "n" \neq 0) (Stmt.assign "n" (fun s \mapsto s "n" - 1); Stmt.assign "m" (fun s \mapsto s "m" + 1))) {* fun s \mapsto s "n" = 0 \land s "m" = n_0 + m_0 *} from by vcg <;> aesop
```

First, we use show to annotate the while loop with an invariant. Recall that the show command restates the goal in a computationally equivalent way. Here, we use this facility to replace Stmt.whileDo by Stmt.invWhileDo, which equals Stmt.whileDo by definition. The program and its pre- and postconditions are otherwise the same as in the theorem statement.

We invoke vcg as the first proof step. This will apply all the necessary Hoare rules and leave us with some subgoals, which are no match for aesop.

# 10.8 Hoare Triples for Total Correctness

So far this chapter's focus has been on partial correctness. When we state the Hoare triple {P} S {Q}, we claim that the final state will satisfy Q if the program S terminates, but we say nothing when S does not terminate. In particular, we can prove any postcondition for the diverging program while True do skip. This is admittedly too liberal: If you are asked to program a sorting algorithm at an exam, you should certainly not give while True do skip as your answer.

Total correctness is a stronger notion that asserts, in addition to partial correctness, that the program terminates normally. We first focused on partial correctness because it is simpler and because it is a necessary component of total correctness anyway.

The Hoare triples for total correctness have the form [P] S [Q], with the following intended meaning:

If the precondition P holds before S is executed, the execution terminates normally and the postcondition Q holds in the final state.

For deterministic programs, this can be expressed equivalently as follows:

If the precondition P holds before S is executed, there exists a state in which execution terminates normally and the postcondition Q holds in that state.

Here is an example Hoare triple that is valid:

[i 
$$\leq$$
 10] while i < 10 do i := i + 1 [i = 10]

For the WHILE language, the distinction between partial and total correctness only concerns while loops (and programs containing them). The Hoare rule for while must now be annotated by a *variant* v—a natural number that decreases by one or more with each iteration:

$$\frac{ [P \land b \land v = v_0] \ S \ [P \land v < v_0] }{ [P] \ while \ b \ do \ S \ [P \land \neg b] } WHILE-VAR$$

Here,  $v_0$  is a logical variable that freezes v's initial value and whose scope is the entire premise, whereas v is a mathematical variable (like P, b, and S). For the example above, we could take 10 - i as the variant (or 50 - i or 1024 - i \* i).

Consider an execution  $s_0$ ,  $s_1$ , ...,  $s_{n-1}$ ,  $s_n$  corresponding to n loop iterations, as in Section 10.2. The following conditions will hold:

The theory of Hoare triples for total correctness is developed in this chapter's exercise sheet.

# **Chapter 11**

# **Denotational Semantics**

We now review a third way to specify the semantics of a programming language: denotational semantics. A denotational semantics directly specifies the meaning of programs as a mathematical object. If an operational semantics corresponds to an idealized interpreter and a Hoare logic corresponds to an idealized verifier, then a denotational semantics corresponds to an idealized compiler: a compiler that translates the source program not to assembly language but directly to mathematics.

The core of this chapter is modeled closely after Chapter 11 of Concrete Semantics: With Isabelle/HOL [23].

# 11.1 Compositionality

A denotational semantics defines the meaning of each program as a mathematical object. Abstractly, it can be viewed as a function

$$[\![\,]\!]: \mathsf{syntax} \to \mathsf{semantics}$$

A key property of denotational semantics is *compositionality*: The meaning of a compound statement should be defined in terms of the meaning of its components. Consider the straightforward definition

$$[S] = \{(s, t) \mid (S, s) \Longrightarrow t\}$$

in terms of the big-step semantics  $(\Longrightarrow)$ . Actually, because Lean supports only variables on the left-hand side of the vertical bar, we must write

$$[S] = \{st \mid (S, Prod.fst st) \Longrightarrow Prod.snd st\}$$

This definition specifies the desired semantics, but it does not qualify as a denotational semantics due to lack of compositionality: The meaning of the compound statements (sequential composition, if-then-else, and while) is given directly, without using the denotation of their components.

A fully compositional definition allows us to reason equationally about programs, which is often more convenient than using the introduction, elimination, and inversion principles of  $\implies$ . In essence, we want structurally recursive equations of the form

with no occurrences of S and T in the right-hand sides except as arguments to  $[\![\,]\!]$ . An evaluation function for arithmetic expressions

$$\texttt{eval}: \underbrace{\texttt{AExp}}_{\texttt{syntax}} \to \underbrace{(\texttt{String} \to \mathbb{Z}) \to \mathbb{Z}}_{\texttt{semantics}}$$

satisfying equations such as

```
eval (AExp.add e_1 e_2) env = eval e_1 env + eval e_2 env
```

is a denotational semantics, because the semantics of AExp.add  $e_1$   $e_2$  is defined in terms of the semantics of  $e_1$  and of  $e_2$ .

Denotational semantics are a natural match for arithmetic expressions and also for functional programs. Now we want a convenient denotational semantics for imperative programs. Because of while loops, which are not guaranteed to terminate, we need to develop some additional mathematical notions to reach the point where we can formulate the desired semantics.

#### 11.2 A Relational Denotational Semantics

Denotational semantics for deterministic languages are normally given as a function from prestate to poststate, but relations are more general and more convenient to manipulate. We present a relational denotational semantics.

A denotational semantics of a program will be a mathematical object of type Set (State  $\times$  State). The relational approach was also used for the big-step semantics, which took the form of a predicate of type State  $\rightarrow$  State  $\rightarrow$  Prop. These two types are isomorphic, but Set  $\alpha$ , defined as  $\alpha \rightarrow$  Prop, supports many useful operations and notations, such as  $\emptyset$ ,  $\cup$ ,  $\cap$ ,  $\in$ , and  $\{\ldots \mid \ldots\}$  (Section 7.7).

Our semantics is called denote, with [] as syntactic sugar. We start with the first four equations of the definition, while keeping while for later:

The skip statement is interpreted as the identity relation over states—i.e., the set of all tuples of the form (s, s). This captures the desired semantics of skip: The poststate is always identical to the prestate.

The semantics of assignment is the set of tuples where the second component reflects the result of the assignment.

The semantics of sequential composition is elegantly expressed as a relational composition  $\bigcirc$ , which is defined by the equation

```
r_1 \bigcirc r_2 = \{ac \mid \exists b, (Prod.fst ac, b) \in r_1 \land (b, Prod.snd ac) \in r_2\}
```

11.3. Fixpoints

The definition is perhaps easier to relate to when expressed as

$$r_1 \bigcirc r_2 = \{(a, c) \mid \exists b, (a, b) \in r_1 \land (b, c) \in r_2\}$$

Unfortunately, Lean cannot process that form, since it requires the left part of a comprehension to be a single variable.

The semantics of an if-then-else statement is given as the union of the semantics of the two branches, restricted to include only the tuples whose first component makes the Boolean condition true or false, depending on the branch. The restriction operator is defined by

```
r \mid P = \{ab \mid ab \in r \land P (Prod.fst ab)\}
```

A more intuitive definition would have been

```
r \mid P = \{(a, b) \mid (a, b) \in r \land P a\}
```

but again this is not supported by Lean. Note two degenerate cases: If  $P := (fun \rightarrow True)$ , then  $r \mid P = r$ , and if  $P := (fun \rightarrow False)$ , then  $r \mid P = \emptyset$ .

The difficulties arise when we try to define the semantics of while loops. We would like to write

```
| Stmt.whileDo B S =>
((denote S ○ denote (Stmt.whileDo B S)) ↓ B)
∪ (Id ∫ (fun s ↦ ¬ B s))
```

but this is ill-founded due to the recursive call on Stmt.whileDo B S. We need something else. What we are looking for on the right-hand side is some term X that satisfies the equation

```
X = ((denote S \bigcirc X) \mid B) \cup (Id \mid (fun s \mapsto \neg B s))
```

We are looking for what mathematicians call a fixpoint. The next four sections are concerned with building an operator lfp that computes the fixpoint for a given equation. Using lfp, we will be able to define the semantics of while loops by

```
| Stmt.whileDo B S => lfp (fun X \mapsto ((denote S \bigcirc X) \downarrow B) \cup (Id \downarrow (fun s \mapsto \neg B s)))
```

# 11.3 Fixpoints

A fixpoint (or fixed point) of f is a solution for X in the equation

$$X = f X$$

In general, fixpoints may not exist at all for some f; for example, if f := Nat.succ, then there exists no value X such that X = Nat.succ X. There may also be several fixpoints; for example, if  $f := (fun \ x \mapsto x)$ , then any X is a solution of  $X = (fun \ x \mapsto x) \ X = X$ . Under some conditions on f, a unique *least fixpoint* and a unique *greatest fixpoint* are guaranteed to exist.

Consider the following fixpoint equation, where  $X : \mathbb{N} \to \mathsf{Prop}$ :

$$X = (fun n : \mathbb{N} \mapsto n = 0 \lor (\exists m : \mathbb{N}, n = m + 2 \land X m))$$

This equation is the  $\beta$ -reduced variant of an equation with the right format:

$$X = \overbrace{\left(\text{fun}\left(P : \mathbb{N} \to \text{Prop}\right)\left(n : \mathbb{N}\right) \mapsto n = 0 \lor \left(\exists m : \mathbb{N}, n = m + 2 \land P m\right)\right)}^{\text{f}} X$$

A solution is X := Even, the predicate that characterizes the even natural numbers. Recall that we proved the inversion rule

Even 
$$n \leftrightarrow n = 0 \lor (\exists m : \mathbb{N}, n = m + 2 \land \text{Even m})$$

in Section 6.5. It turns out that Even is the only fixpoint. In general, the least and greatest fixpoint may be different. Consider the equation

$$X = (fun P \mapsto P) X$$

for X :  $\mathbb{N} \to \mathsf{Prop}$ . The least fixpoint is fun  $\_ \mapsto \mathsf{False}$  and the greatest fixpoint is fun  $\_ \mapsto \mathsf{True}$ . By convention, we have  $\mathsf{False} < \mathsf{True}$  and thus (fun  $\_ \mapsto \mathsf{False}$ ) < (fun  $\_ \mapsto \mathsf{True}$ ). Similarly,  $\emptyset < \mathsf{@Set.univ}\ \alpha$  for any inhabited type  $\alpha$ .

For the semantics of while loops, X will have type Set ( $State \times State$ ) of relations between states, and f will correspond to either taking one extra iteration of the loop (if the condition B is true) or the identity (if B is false).

Which fixpoint should we use for the semantics of while? Whereas the greatest fixpoint would also allow cyclic and diverging executions, the least fixpoint allows only finite (but possibly unbounded) executions. Hence we choose the least fixpoint.

# 11.4 Monotone Functions

We claimed above that the least and greatest fixpoints are guaranteed to exist under some conditions on f. It is time to make this more precise. Let  $\alpha$  and  $\beta$  be arbitrary types, each equipped with a partial order  $\leq$ . The function  $f: \alpha \to \beta$  is monotone if  $a \leq b \to f$  a  $\leq f$  b for all a, b. The function  $f: Set \alpha \to Set \alpha$  admits least and greatest fixpoints if f is monotone.

Many operations on sets (e.g., union  $\cup$ ), relations (e.g., composition  $\bigcirc$ ), and functions (e.g., the identity function  $\operatorname{fun} x \mapsto x$ , the constant function  $\operatorname{fun} \_ \mapsto k$ , composition  $\circ$ ) are monotone or preserve monotonicity. Of course, not all functions are monotone. Here is an example of a nonmonotone function f on  $\operatorname{Set} \alpha$ , with  $\subseteq$  as the partial order:

$$f A = (if A = \emptyset then Set.univ else \emptyset)$$

If  $\alpha$  is inhabited, we have  $\emptyset \subseteq \mathsf{Set.univ}$  but  $\mathsf{f} \emptyset = \mathsf{Set.univ} \not\subseteq \emptyset = \mathsf{f} \mathsf{Set.univ}$ .

# 11.5 Complete Lattices

To define lfp for sets (including relations), we need two operations: subset  $\subseteq$ : Set  $\alpha \to \operatorname{Set} \alpha \to \operatorname{Prop}$  and big intersection  $\cap$ : Set (Set  $\alpha$ )  $\to$  Set  $\alpha$ , which can be defined as

$$\bigcap X = \{a \mid \forall A, A \in X \rightarrow a \in A\}$$

If X is a finite set  $\{A_1, \ldots, A_n\}$ , then  $\bigcap X = A_1 \cap \cdots \cap A_n$ .

We can define lfp more generally so that it works not only with sets but with any instance of the algebraic structure called a complete lattice. A complete lattice  $\alpha$  is an ordered type for which each Set  $\alpha$  has an infimum, also called *greatest lower bound*. A complete lattice consists of

- 1. a partial order  $\leq$  :  $\alpha \rightarrow \alpha \rightarrow \text{Prop}$  (i.e., a reflexive, antisymmetric, and transitive binary predicate);
- 2. an operator  $\square$ : Set  $\alpha \to \alpha$ , called the *infimum*.

The  $\square$  operator satisfies the following two conditions:

- 1.  $\square$  A is a lower bound of A:  $\square$  A  $\leq$  b for all b  $\in$  A;
- 2.  $\square$  A is a greatest lower bound:  $b \leq \square$  A for all b such that  $\forall a, a \in A \rightarrow b \leq a$ .

Together, conditions 1 and 2 ensure that  $\prod$  A is the unique greatest lower bound.

The lattice operators  $\leq$  and  $\prod$  generalize  $\subseteq$ : Set  $\alpha \to \text{Set } \alpha \to \text{Prop and } \cap$ : Set (Set  $\alpha$ )  $\to$  Set  $\alpha$ . Be aware that  $\prod$  A need not be in A. For example, an open interval ]a,b[ over  $\mathbb R$  has infimum  $a \notin ]a,b[$ . The infimum of a set is a generalization of the notion of minimal element.

Here are some examples of complete lattices:

- Set  $\alpha$  with respect to  $\subseteq$  and  $\bigcap$  for all types  $\alpha$ ;
- Prop with respect to  $\rightarrow$  and fun A  $\mapsto \forall$ a  $\in$  A, a;
- ENat :=  $\mathbb{N} \cup \{\infty\}$  with respect to  $\leq$  and a suitable infimum operator;
- EReal :=  $\mathbb{R} \cup \{-\infty, \infty\}$  with respect to  $\leq$  and a suitable infimum operator.

If  $\alpha$  is a complete lattice, then  $\beta \to \alpha$  is also a complete lattice. If  $\alpha$  and  $\beta$  are complete lattices, then  $\alpha \times \beta$  is also a complete lattice. In both cases,  $\leq$  and  $\square$  are defined componentwise.

Here are some nonexamples of complete lattices:  $\mathbb{N}$ ,  $\mathbb{Z}$ ,  $\mathbb{Q}$ , and  $\mathbb{R}$  with respect to  $\leq$ . The issue is that there is no greatest element to assign to  $\bigcap \emptyset$ . Another nonexample is ERat :=  $\mathbb{Q} \cup \{-\infty, \infty\}$ , because  $\bigcap \{q \mid 2 < q * q\} = sqrt 2$  is not in ERat.

In Lean, it is natural to represent complete lattices using type classes:

```
class CompleteLattice (\alpha: Type)
extends PartialOrder \alpha: Type where
Inf : Set \alpha \to \alpha
Inf_le : \forall A b, b \in A \to Inf A \leq b
le_Inf : \forall A b, (\forall a, a \in A \to b \leq a) \to b \leq Inf A \in A
```

The type Set  $\alpha$  is an instance of the type class:

```
instance Set.CompleteLattice \{\alpha: \mathsf{Type}\}: CompleteLattice (Set \alpha) := \{ @Set.PartialOrder \alpha with Inf := fun \mathsf{X} \mapsto \{\mathsf{a} \mid \forall \mathsf{A}, \; \mathsf{A} \in \mathsf{X} \to \mathsf{a} \in \mathsf{A}\} Inf_le := by aesop le_Inf := by aesop \}
```

# 11.6 Least Fixpoint

Using complete lattices, we can define the least fixpoint operator:

$$lfp f = \prod \{x \mid f x \le x\}$$

In Lean:

```
\begin{array}{lll} \text{def lfp } \{\alpha: \text{Type}\} \text{ [CompleteLattice } \alpha] \text{ (f : } \alpha \to \alpha): \alpha: = \\ \text{CompleteLattice.Inf } \{a \mid f \mid a \leq a\} \end{array}
```

The Knaster-Tarski theorem,<sup>1</sup> which we briefly mentioned in Section 6.1, gives us the following properties for any monotone function f on a complete lattice:

- lfp f is a fixpoint: lfp f = f (lfp f);
- lfp f is smaller than any other fixpoint:  $X = f X \rightarrow lfp f \leq X$ .

# 11.7 A Relational Denotational Semantics, Continued

With lfp, we can fulfill our promise and complete the definition of the denotational semantics of WHILE programs:

```
| Stmt.whileDo B S => lfp (fun X \mapsto ((denote S \bigcirc X) \downarrow B) \cup (Id \downarrow (fun s \mapsto \neg B s)))
```

To validate our definition, we can prove the following connection between the denotational and the big-step semantics:

```
theorem denote_Iff_BigStep (S : Stmt) (s t : State) : (s, t) \in [S] \leftrightarrow (S, s) \Longrightarrow t
```

For the proof, we refer to Chapter 11 of *Concrete Semantics: With Isabelle/HOL* [23] or to the demonstration file accompanying this chapter.

# 11.8 Application to Program Equivalence

Based on the denotational semantics, we introduce a notion of program equivalence. Two programs are equivalent if they have the same semantics:

```
def DenoteEquiv (S_1 S_2 : Stmt) : Prop := ||S_1|| = ||S_2||
```

We write  $S_1 \sim S_2$  as an abbreviation for DenoteEquiv  $S_1 S_2$ . It is easy to see that  $\sim$  is an equivalence relation.

Program equivalence can be used to replace a subprogram in a larger program with another subprogram if they have the same semantics. This is achieved by the following congruence rules:

```
theorem DenoteEquiv.seq_congr \{S_1 \ S_2 \ T_1 \ T_2 : Stmt\}

(hS : S_1 \sim S_2) (hT : T_1 \sim T_2) :

S_1; T_1 \sim S_2; T_2 :=

by
```

<sup>&</sup>lt;sup>1</sup>https://en.wikipedia.org/wiki/Knaster-Tarski theorem

```
simp [DenoteEquiv, denote] at *
simp [*]

theorem DenoteEquiv.if_congr {B} {S<sub>1</sub> S<sub>2</sub> T<sub>1</sub> T<sub>2</sub> : Stmt}
    (hS : S<sub>1</sub> ~ S<sub>2</sub>) (hT : T<sub>1</sub> ~ T<sub>2</sub>) :
Stmt.ifThenElse B S<sub>1</sub> T<sub>1</sub> ~ Stmt.ifThenElse B S<sub>2</sub> T<sub>2</sub> :=
by
    simp [DenoteEquiv, denote] at *
    simp [*]

theorem DenoteEquiv.while_congr {B} {S<sub>1</sub> S<sub>2</sub> : Stmt}
    (hS : S<sub>1</sub> ~ S<sub>2</sub>) :
Stmt.whileDo B S<sub>1</sub> ~ Stmt.whileDo B S<sub>2</sub> :=
by
    simp [DenoteEquiv, denote] at *
    simp [*]
```

A congruence rule is a theorem that lifts an equivalence relation over some context (here, ~ over sequential composition, if-then-else, and while).

Notice how the denotational semantics leads to short proofs by rewriting. This should not be surprising, given that it is designed to be equational and compositional. If we had used the big-step semantics as the basis for program equivalence instead, these proofs would have been much more complicated.

We can now prove simple programs equivalent by equational reasoning:

```
theorem DenoteEquiv.skip_assign_id {x}:
   Stmt.assign x (fun s → s x) ~ Stmt.skip :=
   by simp [DenoteEquiv, denote, Id]

theorem DenoteEquiv.seq_skip_left {S}:
   Stmt.skip; S ~ S :=
   by simp [DenoteEquiv, denote, Id, comp]

theorem DenoteEquiv.seq_skip_right {S}:
   S; Stmt.skip ~ S :=
   by simp [DenoteEquiv, denote, Id, comp]
```

We defined the semantics of while using the lfp operator, but who knows whether monotonicity—which guarantees the existence of a least fixpoint—is met? To quell such doubts, we prove the following theorem:

```
theorem DenoteEquiv.if_seq_while {B S} :
   Stmt.ifThenElse B (S; Stmt.whileDo B S) Stmt.skip
   ~ Stmt.whileDo B S :=
   by
     simp [DenoteEquiv, denote]
   apply Eq.symm
   apply lfp_eq
   apply Monotone_while_lfp_arg
```

The theorem gives us a convenient way to expand or contract one iteration of a loop. The second apply invokes the theorem lfp\_eq: lfp f = f (lfp f), stating

that lfp is a fixpoint. The last step applies a theorem to convince Lean that lfp's argument is monotone. That theorem's proof is rather monotonous:

```
theorem Monotone_while_lfp_arg (S B) :
   Monotone (fun X → [S] ○ X ↓ B ∪ Id ↓ (fun s → ¬ B s)) :=
   by
    apply Monotone_union
   { apply SorryTheorems.Monotone_restrict
        apply SorryTheorems.Monotone_comp
        { exact Monotone_const _ }
        { exact Monotone_id } }
        { apply SorryTheorems.Monotone_restrict
        exact Monotone_id } }
```

# 11.9 A Simpler Approach Based on an Inductive Predicate

Lean's inductive predicates correspond to least fixpoints, but they are built into Lean's logic (the calculus of inductive constructions), without using a fixpoint operator like lfp. Most of this section was dedicated to the construction of an lfp operator. Could we have used an inductive predicate instead and saved ourselves this work?

The answer is yes. First, recall that the types Set ( $State \times State$ ) and  $State \rightarrow State \rightarrow Prop$  are equivalent representations of binary relations in Lean. So we can build the following Awhile predicate, which resembles the reflexive transitive closure but stops when the given condition B turns false:

```
inductive Awhile (B : State → Prop)
   (r : Set (State × State)) :
   State → State → Prop
   | true {s t u} (hcond : B s) (hbody : (s, t) ∈ r)
        (hrest : Awhile B r t u) :
        Awhile B r s u
   | false {s} (hcond : ¬ B s) :
        Awhile B r s s
```

With this operator, the definition of the definitional semantics becomes

The Awhile predicate's introduction rules closely resemble the while rules of the big-step operational semantics. Behind their different façades, operational and denotational semantics are not so dissimilar after all.

# Part IV Mathematics

# **Chapter 12**

# Logical Foundations of Mathematics

In this chapter, we dive deeper into the logical foundations of Lean. Most of the features described here are especially relevant for defining mathematical objects and proving theorems about them. For even more details, we refer to Carneiro's MSc thesis [5].

#### 12.1 Universes

In dependent type theory, not only do all terms have a type, but also all types have types themselves. We have already seen some occurrences of this. The PAT principle tells us to view proofs as terms and propositions as types. For example, the theorem

```
\operatorname{@And.intro}: \forall a \ b, \ a \to b \to a \wedge b
```

is really a term @And.intro of type  $\forall a \ b$ ,  $a \rightarrow b \rightarrow a \land b$ , which in turn has a type:

```
\forall a \ b, \ a \rightarrow b \rightarrow a \land b : Prop
```

What is the type of Prop then? Prop has the same type as virtually all other types we have constructed so far:

```
Prop : Type
```

What is the type of Type? The simplest solution would be to let Type: Type, but this choice leads to Girard's paradox, the type theory equivalent of Russel's paradox. To avoid inconsistencies, we need a fresh type to contain Type, which we call Type 1. Type 1 itself has type Type 2, and so on:

```
Type : Type 1
Type 1 : Type 2
Type 2 : Type 3
:
```

In fact, Type with no argument is an abbreviation for Type o. If we want to incorporate Prop in this hierarchy, we can use the syntax Sort u, where Sort o is

an alias for Prop and Sort (u + 1) is an alias for Type u. The hierarchy is captured by the following typing judgment:

All of these types containing other types are called *universes*, and the u in the expression Sort u is a *universe level*. Although universe levels look like terms of type  $\mathbb{N}$ , in fact they are not even terms.

Instead of using Type everywhere, you can make your theorems slightly more general without having to think about universe levels by writing Type \_. Lean then creates a fresh universe variable. This helps maintain the illusion that we work in a convenient logic where Type: Type holds, but without introducing any paradoxes. In practice, Type o is large enough for most of computer science and mathematics.

# 12.2 The Peculiarities of Prop

Although Prop may seem to fit nicely into the universe hierarchy, it differs from the other universes in several respects.

## 12.2.1 Impredicativity

When constructing a new type from other types (e.g.,  $\alpha \to \beta$  from  $\alpha$ : Type u and  $\beta$ : Type v), the newly constructed type is more complex than each of its components, and it is natural to put it into the largest universe involved (e.g.,  $\alpha \to \beta$ : Type (max u v)). This is exactly what Lean does. The following typing rule expresses this idea generally for dependent types:

$$\frac{\mathsf{C} \vdash \sigma : \mathsf{Type} \; \mathsf{u} \qquad \mathsf{C,} \; \mathsf{x} : \sigma \vdash \tau[\mathsf{x}] : \mathsf{Type} \; \mathsf{v}}{\mathsf{C} \vdash (\mathsf{x} : \sigma) \to \tau[\mathsf{x}] : \mathsf{Type} \; (\mathsf{max} \; \mathsf{u} \; \mathsf{v})} \mathsf{ARROW-TYPE}$$

This behavior of the Type universes is called *predicativity*. In general, predicativity means that an object *must not* be defined in terms of a quantifier ranging over that same object.

However, it is convenient to have Prop behave differently. We would like the expression  $\forall a: Prop$ ,  $a \to a$  to be of type Prop—it is, after all, a proposition. Unfolding the syntactic sugar for  $\forall$ , this expression is the same as  $(a: Prop) \to a \to a$ . If we had Type u instead of Prop, the above typing rule would yield

$$(a : Type u) \rightarrow a \rightarrow a : Type (u + 1)$$

since Type u: Type (u+1) and max(u+1) (max u u) = u+1. Thus, the universe level is increased by one when typing this expression. To force expressions such as  $\forall a:$  Prop,  $a \rightarrow a$  to have type Prop anyway, we need a special typing rule for  $\forall$  expressions with a Prop body:

$$\frac{\mathsf{C} \vdash \sigma : \mathsf{Sort} \; \mathsf{u} \qquad \mathsf{C}, \; \mathsf{x} : \sigma \vdash \tau[\mathsf{x}] : \mathsf{Prop}}{\mathsf{C} \vdash (\forall \mathsf{x} : \sigma, \tau[\mathsf{x}]) : \mathsf{Prop}} \mathsf{ARROW-PROP}$$

The rule yields

```
\forall a : Prop, a \rightarrow a : Prop
```

as desired. The two typing rules above can be summarized as the single rule

```
\frac{\mathsf{C} \vdash \sigma : \mathsf{Sort} \; \mathsf{u} \qquad \mathsf{C}, \; \mathsf{x} : \sigma \vdash \tau[\mathsf{x}] : \mathsf{Sort} \; \mathsf{v}}{\mathsf{C} \vdash (\mathsf{x} : \sigma) \rightarrow \tau[\mathsf{x}] : \mathsf{Sort} \; (\mathsf{imax} \; \mathsf{u} \; \mathsf{v})} \mathsf{ARROW}
```

where imax u o = o and imax u (v + 1) = max u (v + 1). This behavior is called *impredicativity* of Prop. In general, impredicativity means that an object *may* be defined via a quantifier ranging over itself.

#### 12.2.2 Proof Irrelevance

A second difference between Prop and Type is *proof irrelevance*. It means that any two proofs of the same proposition a are equal:

```
theorem proof_irrel {a : Prop} (h<sub>1</sub> h<sub>2</sub> : a) :
    h<sub>1</sub> = h<sub>2</sub> :=
    by rfl
```

In Lean, this equality is a syntactic equality up to computation, allowing us to use the rfl tactic. When viewing a proposition as a type and a proof as an element of that type, proof irrelevance means that a proposition is either an empty type or has exactly one inhabitant. If it is empty, it is false. If it has exactly one inhabitant, it is true. Proof irrelevance is very helpful when reasoning about dependent types.

In Section 6.3, we saw a diagram depicting the interpretation of Bool and Prop side by side. The diagram did not take proof irrelevance into account and showed multiple proofs for the same proposition. We now know this is not accurate. Here is a revised diagram:



If we mentally connect Bool's two elements (false and true) with Prop's two types (False and True), we see that Prop is nearly the same as Bool, except that propositions of type Prop can store proofs, in accordance with the PAT principle. The following table summarizes the situation:

False: Prop
True.intro: True: Prop
false: Bool: Type
true: Bool: Type

To make proof irrelevance work, Lean must give up the "no confusion" property for inductive predicates. Indeed, various proofs of the same proposition should definitely be "confused." This only concerns inductive predicates (e.g., Even), not inductive types in general (e.g., List  $\alpha$ ).

Other systems and logics make different choices. For example, Coq is proof-relevant by default but compatible with proof irrelevance. Homotopy type theory and other constructive or intuitionistic type theories build on data in equality proofs and are therefore incompatible with proof irrelevance.

#### 12.2.3 No Large Elimination

A further difference between Prop and Type is that Prop does not allow large elimination: It is generally impossible to extract information from a proof of a proposition and use it in a program (i.e., a value of a type belonging to Type). After all, because of proof irrelevance, all proofs of a given proposition are equal, so they cannot carry specific information that would distinguish them.

Imagine we could extract information from proofs inside a program. We could for instance use the match construct in function definitions such as the following:

```
-- fails
def unsquare (i : Z) (hsq : ∃j, i = j * j) : Z :=
  match hsq with
  | Exists.intro j _ => j
```

The unsquare function takes a square number i and a proof hsq that i is actually a square number and returns the number j before squaring, extracted from the proof. Lean raises the error

```
tactic 'induction' failed, recursor 'Exists.casesOn' can only eliminate into Prop
```

If it accepted the definition, we could derive False as follows. Let

```
hsq<sub>1</sub> := Exists.intro 3 (by linarith)
hsq<sub>2</sub> := Exists.intro (-3) (by linarith)
```

be two proofs of  $\exists j$ ,  $(9:\mathbb{Z}) = j*j$ . Notice that they use different witnesses for j (3 versus -3). We then have unsquare  $9 \text{ hsq}_1 = 3$  and unsquare  $9 \text{ hsq}_2 = -3$ . Yet, by proof irrelevance,  $\text{hsq}_1 = \text{hsq}_2$ . Hence, unsquare  $9 \text{ hsq}_2$  equals 3. But we already determined that it equals -3. This means 3 = -3, a contradiction.

An unfortunate consequence of the absence of large elimination is that we cannot perform rule induction by pattern matching and recursion. This kind of induction relies on a "measure"—a function to  $\mathbb N$  that assigns a size to the arguments. Without large elimination, the measure cannot be defined meaningfully. This explains why we always use the induction tactic for rule induction.

As a compromise, Lean allows *small elimination*, which eliminates only into Prop—whereas large elimination can eliminate into an arbitrary large universe Type u. This means we can use match to analyze the structure of a proof, extract existential witnesses, and so on, as long as the match expression itself is in a proof. We have seen several examples of this in Section 6.5.

As a further compromise, Lean allows large elimination for syntactic subsingletons: types in Prop that can be proved in at most one way. For example, False has no proof, and all proofs of a  $\land$  b have the form And.intro \_ \_. (Recursively, there might be multiple ways to prove a and b.) More precisely, a syntactic subsingleton is an inductive definition with at most one constructor whose arguments are either Prop or appear as immediate arguments in the result type. When we match h: a  $\land$  b against the pattern And.intro ha hb, no information about h is leaked.

# 12.3 The Axiom of Choice

Lean's logic includes the axiom of choice, which makes it possible to obtain an arbitrary element of any nonempty type. Consider the following predicate:

```
inductive Nonempty (\alpha : Sort u) : Prop
| intro (val : \alpha) : Nonempty \alpha
```

The predicate states that  $\alpha$  has at least one element. To prove Nonempty  $\alpha$ , we must provide an  $\alpha$  value to Nonempty.intro:

```
theorem Nat.Nonempty :
  Nonempty N :=
  Nonempty.intro ⊙
```

Since Nonempty lives in Prop, large elimination is not available, and thus we cannot extract the element that was used in the proof of Nonempty  $\alpha$ .

In Lean, the axiom of choice takes the form of a function that returns an arbitrary  $\alpha$  value given a proof of Nonempty  $\alpha$ :

```
Classical.choice \{\alpha : Sort u\} : Nonempty \alpha \rightarrow \alpha
```

We have no way to know whether the returned element is the same element that was used to prove Nonempty  $\alpha$ . It will just be an arbitrary element of  $\alpha$ .

The constant Classical.choice is noncomputable. If we ask Lean for its value using #reduce or #eval, it will refuse to compute it. In other words, proofs may be terms, but they are not necessarily programs. This is one of the reasons why some logicians prefer to work without the axiom of choice. By contrast, the vast majority of mathematicians have no objections against the axiom.

Unlike proof irrelevance and large and small elimination, the axiom of choice is not built into the Lean kernel; it is only an axiom in the core library, and we are free to work without it. Lean requires us to mark definitions with the noncomputable keyword if they use Classical.choice to define constants in Type.

The following tools rely on Classical.choice:

■ The function Classical.choose, often called Hilbert's choice operator, can help us find a witness of  $\exists a:\alpha$ , p a if we do not care which one. Its companion Classical.choose\_spec gives a proof that the witness is indeed a witness.

```
Classical.choose : (\exists a:\alpha,p\;a)\to\alpha Classical.choose_spec : \forall h:(\exists a:\alpha,p\;a), p (Classical.choose h)
```

Intuitively, the choice operator tells us, "Convince me that there exists an element satisfying p, and I will give you such an element."

• We can also derive the traditional axiom of choice:

```
Classical.axiomOfChoice (\alpha \beta : Type) {R : \alpha \rightarrow \beta \rightarrow Prop} : (\forall x : \alpha, \exists y : \beta, R \times y) \rightarrow (\exists f, \forall x, R \times (f \times x))
```

From the axiom of choice and propositional and functional extensionality (propext, funext), we can derive the law of excluded middle:

```
Classical.em: \forall a : Prop, a \lor \neg a
```

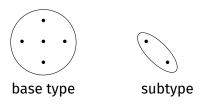
With the law of excluded middle, every proposition is decidable. This means that we can construct proofs based on a case distinction on whether a certain proposition is true.

# 12.4 Subtypes

Inductive types are a very convenient definitional mechanism when they are applicable, but lots of mathematical constructions do not fit that mold. Lean provides two alternatives to cater for these: subtypes and quotients.

Subtyping is a mechanism to create new types from existing ones. Given a predicate on the elements of a base type, the *subtype* contains only those elements of the base type that satisfy the predicate. More precisely, the subtype contains element–proof pairs that combine an element of the base type and a proof that the predicate is true for that element.

The following diagram depicts a subtype that was created by keeping two of the base type's five elements:



Subtyping is useful for those types that cannot be defined as an inductive type. For example, any attempt at defining the type of finite sets along the lines of

```
inductive Finset (\alpha : Type) : Type

| empty : Finset \alpha

| insert : \alpha \rightarrow Finset \alpha \rightarrow Finset \alpha
```

is doomed, because a given set may have multiple representations. For example, {1, 2} can be represented in any of the following ways, and more:

```
Finset.insert 1 (Finset.insert 2 Finset.empty)
Finset.insert 2 (Finset.insert 1 Finset.empty)
Finset.insert 1 (Finset.insert 2 Finset.empty))
```

Instead, we can define finite sets as the subtype of (possibly infinite) sets that are finite. In general, subtypes have the syntax

```
{variable: base-type // property-applied-to-variable}
```

We saw an example in Section 4.6, namely,  $\{i: \mathbb{N} \ / \ i \leq n\}$ , which consists of the natural numbers i fulfilling  $i \leq n$ , where n is fixed in the context. The base type is  $\mathbb{N}$ , and the property is fun  $i \mapsto i \leq n$ . A less suggestive but perhaps less confusing syntax for the same type is  $\mathfrak{A}$  Subtype  $\mathbb{N}$  (fun  $i \mapsto i \leq n$ ). Our motivating example, the type of finite sets over some type  $\alpha$ , is specified as  $\{A: \text{Set } \alpha \ / \ \text{Set.Finite} \}$ , where Set.Finite is True if and only if its argument is finite.

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## 12.4.1 First Example: Full Binary Trees

To illustrate subtypes, we will define a type of full binary trees building on the BinTree type from Section 5.8. In Section 6.6.3, we introduced a predicate IsFull that is true if each node of a tree has either zero or two child nodes. Based on this type and this predicate, we can construct a subtype FullBinTree, containing only full binary trees, as follows:

```
\begin{array}{l} \text{def FullBinTree } (\alpha: \text{Type}): \text{Type} := \\ \{\text{t}: \text{BinTree } \alpha \text{ // IsFull t} \} \\ \\ \text{This is syntactic sugar for} \\ \text{def FullBinTree } (\alpha: \text{Type}): \text{Type} := \\ \text{@Subtype } (\text{BinTree } \alpha) \text{ IsFull} \\ \\ \text{where Subtype is defined as follows:} \\ \text{inductive Subtype } \{\alpha: \text{Type}\} \text{ (p}: \alpha \rightarrow \text{Prop)}: \text{Type} \\ \text{| mk}: (x: \alpha) \rightarrow \text{p} \text{ x} \rightarrow \text{Subtype p} \end{array}
```

Elements of FullBinTree are essentially dependently typed pairs, where the first component is a tree t and the second component is a proof that t is full. For example, here are the definitions of an empty full binary tree and a full binary tree consisting of a single inner node labeled 6:

Given a value of type FullBinTree, we can retrieve its two components via Subtype. val and Subtype.property:

```
#reduce Subtype.val fullBinTree6
#check Subtype.property fullBinTree6
```

The most appealing aspect of subtypes is that we can *lift* the operations from the base type to the subtype instead of having to build a library from scratch, as long as they preserve the subtype property. We only need to define "wrappers" around constants from the base type. In general, defining such a wrapper around an operation f on the base type involves three steps:

- 1. extract the base type values from the wrapper arguments;
- 2. invoke f on those base type values;
- 3. encapsulate the result using Subtype.mk, together with a proof that the resulting base type value fulfills the subtype property.

Using this procedure, we can lift BinTree functions to FullBinTree functions if they preserve the property IsFull. For example, to lift the mirror operation

from the type  $\mathtt{BinTree} \to \mathtt{BinTree}$  to the type  $\mathtt{FullBinTree} \to \mathtt{FullBinTree}$  , we must

- extract the BinTree from the wrapper argument;
- invoke mirror on that BinTree;
- 3. encapsulate the result using Subtype.mk, together with a proof that the resulting BinTree fulfills IsFull.

For step 3, we must extract the proof of IsFull from the argument and use the theorem IsFull\_mirror from Section 6.6.3. Putting everything together, we get

```
def FullBinTree.mirror {α : Type} (t : FullBinTree α) :
   FullBinTree α :=
   Subtype.mk (LoVe.mirror (Subtype.val t))
     (by
        apply IsFull_mirror
        apply Subtype.property t)
```

The input is an element t of the subtype FullBinTree. We decompose t into Subtype.val t: BinTree and Subtype.property t: IsFull t. We use the earlier mirror function to reverse the tree component of t and use the theorem IsFull\_mirror, together with the property component of t, to show the condition IsFull (mirror (Subtype.val t)).

Finally, we build a pair containing the resulting tree and the proof that this tree is full using Subtype.mk. The Subtype.mk constructor can be seen both as a pair-like constructor and as a cast from BinTree to FullBinTree, with a second argument that guarantees that the cast is safe.

For proofs about subtypes, the following theorem is useful:

```
Subtype.eq: Subtype.val ?a = Subtype.val ?b \rightarrow ?a = ?b
```

The theorem states that two subtype values are equal if their Subtype.val components are equal. It is crucial that proofs are irrelevant in Lean, because we would not want to have spurious duplicate values in the subtype, differing only in their proofs. Here is how Subtype.eq can be used to prove that the mirror image of the mirror image of a FullBinTree is the FullBinTree itself:

```
theorem FullBinTree.mirror_mirror {α : Type}
  (t : FullBinTree α) :
  (FullBinTree.mirror (FullBinTree.mirror t)) = t :=
  by
    apply Subtype.eq
    simp [FullBinTree.mirror, LoVe.mirror_mirror]
```

Applying the theorem Subtype.eq and unfolding the definition of full\_BinTree. mirror yields the subgoal

```
mirror (mirror (Subtype.val t)) = Subtype.val t
```

which matches the statement of the theorem mirror mirror from Section 5.8.

#### 12.4.2 Second Example: Vectors

As a second example, consider the following definition of vectors:

```
def Vector (\alpha: Type) (n : \mathbb{N}) : Type := {xs : List \alpha // List.length xs = n}
```

A vector is defined as a list of a given length. With lists, there is only one type for a list of all lengths. For vectors, we have one dedicated type for every length of a vector. The advantage of this scheme is that operations such as addition and scalar product of vectors need the two vectors involved to have the same length. We saw a less practical definition of vectors in Section 5.10.

Vectors can be built from lists using Subtype.mk:

```
def vector123 : Vector \mathbb{Z} 3 := Subtype.mk [1, 2, 3] (by rfl)
```

Basic operations on vectors can be defined by decomposing them with Subtype. val and Subtype.property, manipulating the underlying lists, and then recomposing them with Subtype.mk. For example, we can define the componentwise negation of an integer vector as follows:

```
def Vector.neg {n : N} (v : Vector ℤ n) : Vector ℤ n :=
  Subtype.mk (List.map Int.neg (Subtype.val v))
    (by
      rw [List.length_map]
      exact Subtype.property v)
```

We use the function List.map to negate every entry of the underlying list and the theorem List.length\_map to show that this operation does not change the list's length.

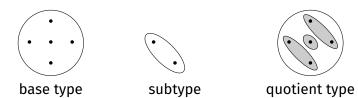
Using Subtype.eq, we can prove the following theorem about Vector.neg:

```
theorem Vector.neg_neg (n : N) (v : Vector Z n) :
  Vector.neg (Vector.neg v) = v :=
  by
    apply Subtype.eq
    simp [Vector.neg]
```

The application of Subtype.eq reduces the goal to showing the corresponding property on the underlying lists. We can then use simp to finish the proof.

## 12.5 Quotient Types

Quotients are a powerful construction in mathematics used to define  $\mathbb{Z}$ ,  $\mathbb{R}$ , and many other sets. Lean supports *quotient types*, an analogous mechanism on types. Like subtyping, quotienting constructs a new type from an existing type. Unlike a subtype, a quotient type contains all of the elements of the base type, except that some elements that are different in the base type may be considered equal in the quotient type. In mathematical terms, the quotient type is isomorphic to a partition of the base type. The diagram below shows a quotient type built from a three-way partition:



The depicted quotient type has only three elements, represented by grayed ellipses. Each of these elements corresponds to one or more base type elements.

The prerequisites to construct a quotient type are a base type  $\tau$  and an equivalence relation R:  $\tau \to \tau \to \mathsf{Prop}$  specifying which elements of the base type will be considered equal in the quotient. To construct the quotient type, we first need to prove that R is an equivalence relation on  $\tau$ . A type  $\tau$  equipped with an equivalence relation is called a *setoid*. In Lean, Setoid is a type class. We can declare an instance using the command

where the ellipses stand for missing proofs of the respective properties. In addition, this instance declaration introduces the notation  $a \approx b$  for R a b. More importantly, we can now use the quotient type Quotient  $\tau$ . Setoid.

Every element a :  $\tau$  belongs to some element in Quotient  $\tau$ . Setoid, given by Quotient.mk  $\tau$ . Setoid a, with

```
Quotient.mk \{\alpha : \mathsf{Type}\} \to (\mathsf{s} : \mathsf{Setoid} \ \alpha) \to \alpha \to \mathsf{Quotient} \ \mathsf{s}
```

The expression Quotient.mk  $\tau$ . Setoid a is quite a mouthful. Fortunately, Lean lets us abbreviate it to to [a].

The following axiom guarantees that pairs of elements for which R holds are indeed equal in the quotient type:

```
Quotient.sound {a b : \tau} : a \approx b \rightarrow [a] = [b]
```

A theorem states the converse:

```
Quotient.exact {a b : \tau} : [\![a]\!] = [\![b]\!] \to a \approx b
```

Finally, we can lift functions of type  $\tau \to \upsilon$ , where  $\upsilon$  is some arbitrary type, to Quotient  $\tau$ . Setoid  $\to \upsilon$  using Quotient.lift, which satisfies the following syntactic equality up to computation. Given some  $f:\tau \to \upsilon$  such that  $h:\forall a\;b$ ,  $a\approx b\to f\;a=f\;b$ , we have

```
Quotient.lift f h [a] = f a
```

for all  $a:\tau$ . The argument h is a proof that f is *compatible* with  $\approx$ ; in other words, it does not distinguish between  $\approx$ -equivalent arguments.

#### 12.5.1 First Example: Integers

As an example of a quotient type, we will construct the integers. A convenient approach is to construct a quotient over pairs of natural numbers. The idea is that a pair (p, n) of natural numbers represents the integer p - n. In this way, we can represent all nonnegative integers p by (p, o) and all negative integers - n by (o, n). We also get many more representations of the same integers; for example, (7, o), (8, 1), (9, 2), and (10, 3) all represent the integer 7.

First, we need to register the equivalence relation that we want to use. We want two pairs  $(p_1, n_1)$  and  $(p_2, n_2)$  to be equal when  $p_1 - n_1$  and  $p_2 - n_2$  yield the same integer. However, the condition  $p_1 - n_1 = p_2 - n_2$  does not work because subtraction on  $\mathbb N$  is ill-behaved (e.g., o - 1 = o). Instead, we use the condition  $p_1 + n_2 = p_2 + n_1$ , which relies on addition.

We provide the definition of our equivalence relation, followed by a proof that it is reflexive, symmetric, and transitive:

```
instance Int.Setoid : Setoid (\mathbb{N} \times \mathbb{N}) :=
   { r :=
         fun pn_1 pn_2 : \mathbb{N} \times \mathbb{N} \mapsto
             Prod.fst pn_1 + Prod.snd pn_2 =
             Prod.fst pn<sub>2</sub> + Prod.snd pn<sub>1</sub>
      iseqv :=
         { refl :=
                by
                   intro pn
                   rfl
             symm :=
                by
                   intro pn<sub>1</sub> pn<sub>2</sub> h
                   rw [h]
             trans :=
                by
                   intro pn<sub>1</sub> pn<sub>2</sub> pn<sub>3</sub> h<sub>12</sub> h<sub>23</sub>
                   linarith } }
```

We can now write  $\approx$  for the equivalence relation:

```
theorem Int.Setoid_Iff (pn<sub>1</sub> pn<sub>2</sub> : \mathbb{N} \times \mathbb{N}) : pn<sub>1</sub> \approx pn<sub>2</sub> \leftrightarrow Prod.fst pn<sub>1</sub> + Prod.snd pn<sub>2</sub> = Prod.fst pn<sub>2</sub> + Prod.snd pn<sub>1</sub> := by rfl
```

Using the name int.rel of our Setoid instance that we registered above, we can then define the integers as

```
def Int : Type :=
  Quotient Int.Setoid
```

We can define the integer zero as

```
def Int.zero : Int :=
    [(0, 0)]
```

In fact, any term of the form [(m, m)] represents zero:

```
theorem Int.zero_Eq (m : N) :
   Int.zero = [(m, m)] :=
   by
    rw [Int.zero]
   apply Quotient.sound
   rw [Int.Setoid_Iff]
   simp
```

Next, we define addition on our new integers. To define functions on a quotient type, we cannot simply define them by pattern matching, as for inductive types. Instead, we define the function on the base type first and then *lift* the definition to the quotient. To achieve this, we must prove that the definition of the function f does not depend on the chosen representative of an equivalence class (i.e.,  $a \approx b \rightarrow f \ a = f \ b$ ). The functions Quotient.lift (for unary functions) and Quotient.lift<sub>2</sub> (for binary functions) can be used to lift functions in this way.

Addition can be defined as adding the pairs of natural numbers componentwise. We then need to provide a proof that this can be lifted to a function on the quotient by showing that  $pn_1 \approx pn_1$ ' and  $pn_2 \approx pn_2$ ' imply

Formally:

```
\begin{array}{l} \text{def Int.add} : \text{Int} \rightarrow \text{Int} \rightarrow \text{Int} := \\ \text{Quotient.lift}_2 \\ \text{(fun pn}_1 \text{ pn}_2 : \mathbb{N} \times \mathbb{N} \mapsto \\ \text{[(Prod.fst pn}_1 + \text{Prod.fst pn}_2, \\ \text{Prod.snd pn}_1 + \text{Prod.snd pn}_2)]]) \\ \text{(by} \\ \text{intro pn}_1 \text{ pn}_2 \text{ pn}_1 \text{' pn}_2 \text{' h}_1 \text{ h}_2 \\ \text{apply Quotient.sound} \\ \text{rw [Int.Setoid_Iff] at *} \\ \text{linarith)} \end{array}
```

The resulting function Int.add has the intended behavior:

```
theorem Int.add_Eq (p_1 \ n_1 \ p_2 \ n_2 : \mathbb{N}) : Int.add [(p_1, \ n_1)] \ [(p_2, \ n_2)] = [(p_1 + p_2, \ n_1 + n_2)] := by rfl
```

It would be very convenient if Lean let us enter this theorem as the definition of Int.add in the first place, presumably with the following syntax:

```
-- fails  \frac{\text{def Int.add}}{\text{lnt.add}}: \text{Int} \to \text{Int} \to \text{Int} \\ \mid [\![(p_1,\ n_1)]\!], [\![(p_2,\ n_2)]\!] \Rightarrow [\![(p_1+p_2,\ n_1+n_2)]\!]
```

This would be a nice and intuitive syntax, but without a proof that the definition is compatible with  $\approx$ , we could define nonsensical functions and use them to derive False. For example, we could define

```
-- fails def Int.fst : Int \rightarrow \mathbb{N} | [(p, n)] => p
```

Notice that Int.fst [(0, 0)] = 0 and Int.fst [(1, 1)] = 1. However, since [(0, 0)] = [(1, 1)], we get 0 = 1, a contradiction.

We can use the characteristic theorem Int.add\_Eq to prove other theorems about Int.add, such as

```
theorem Int.add_zero (i : Int) :
   Int.add Int.zero i = i :=
   by
    induction i using Quotient.inductionOn with
   | h pn =>
      cases pn with
   | mk p n => simp [Int.zero, Int.add_Eq]
```

We invoke the induction tactic with Quotient.inductionOn as the induction principle to perform a case distinction on i, replacing i by [pn], where pn is an arbitrary value of the base type  $\mathbb{N} \times \mathbb{N}$ . Then we perform a case distinction on pn, obtaining a pair (p, n). Finally, we simplify the goal, using the definition of Int.zero and the characteristic equation of Int.add.

#### 12.5.2 Second Example: Unordered Pairs

aesop

Unordered pairs are pairs for which no distinction is made between the first and second components. They are usually written  $\{a, b\}$ . We will introduce the type UPair  $\alpha$  of unordered pairs over  $\alpha$  as the quotient of pairs (a, b) with respect to the relation "contains the same elements as":

```
instance UPair.Setoid (\alpha: Type) : Setoid (\alpha \times \alpha) :=
  { r :=
       fun ab_1 ab_2 : \alpha \times \alpha \mapsto
          (\{Prod.fst ab_1, Prod.snd ab_1\} : Set \alpha) =
          ({Prod.fst ab<sub>2</sub>, Prod.snd ab<sub>2</sub>} : Set \alpha)
    iseqv :=
       { refl := by simp
          symm := by aesop
          trans := by aesop } }
  theorem UPair.Setoid_Iff \{\alpha : Type\} (ab_1 ab_2 : \alpha \times \alpha) :
    ab_1 \approx ab_2 \leftrightarrow
     (\{Prod.fst ab_1, Prod.snd ab_1\} : Set \alpha) =
     (\{Prod.fst ab_2, Prod.snd ab_2\} : Set \alpha) :=
    by rfl
  def UPair (\alpha : Type) : Type :=
    Quotient (UPair.Setoid \alpha)
It is easy to prove that our pairs are really unordered:
  theorem UPair.mk_symm \{\alpha : Type\} (a b : \alpha) :
     ([(a, b)] : UPair \alpha) = [(b, a)] :=
       apply Quotient.sound
       rw [UPair.Setoid_Iff]
```

Another representation of unordered pairs is as sets of cardinality 1 or 2. The following operation converts UPair  $\alpha$  values to that representation:

### 12.5.3 Alternative Definitions via Normalization and Subtyping

Each element of a quotient type correspond to a class of  $\approx$ -equivalent elements of the base type. If there exists a systematic way to obtain a canonical representative for each  $\approx$ -equivalence class, we can use a subtype instead of a quotient, keeping only the canonical representatives and filtering out the other elements.

Consider the quotient type Int of integers constructed above. We observed that (7, 0), (8, 1), (9, 2), and (10, 3) all represent the integer 7, but intuitively (7, 0) seems preferable to the others. We will say that a pair (p, n) is canonical if p or n is 0:

```
inductive Int.IsCanonical : \mathbb{N} \times \mathbb{N} \to \mathsf{Prop} | nonpos \{n : \mathbb{N}\} : Int.IsCanonical (0, n) | nonneg \{p : \mathbb{N}\} : Int.IsCanonical (p, 0)
```

The integers then consist of the canonical pairs of natural numbers:

```
def Int : Type := \{pn : \mathbb{N} \times \mathbb{N} // \text{ Int.IsCanonical pn}\}
```

Clearly, each integer can be represented in one and only one way. Operations on integers such as + and \* must then provide canonical results. Fortunately, normalizing pairs of natural numbers is easy:

```
def Int.normalize : \mathbb{N} \times \mathbb{N} \to \mathbb{N} \times \mathbb{N}
  | (p, n) => if p \geq n then (p - n, 0) else (0, n - p)
theorem Int.IsCanonical_normalize (pn : \mathbb{N} \times \mathbb{N}) :
  Int.IsCanonical (Int.normalize pn)
```

For unordered pairs, there is no obvious normal form, except to always put the smaller element first (or last). This requires a linear order  $\leq$  on  $\alpha$ :

```
def UPair.IsCanonical \{\alpha : \mathsf{Type}\}\ [\mathsf{LinearOrder}\ \alpha] : \\ \alpha \times \alpha \to \mathsf{Prop} \\ \mid (\mathsf{a},\ \mathsf{b}) \Rightarrow \mathsf{a} \leq \mathsf{b} \\ \\ \mathsf{def}\ \mathsf{UPair}\ (\alpha : \mathsf{Type})\ [\mathsf{LinearOrder}\ \alpha] : \mathsf{Type} := \\ \{\mathsf{ab}\ : \ \alpha \times \alpha\ //\ \mathsf{UPair}.\mathsf{IsCanonical}\ \mathsf{ab}\} \\ \\
```

Returning to Int.IsCanonical, we note that there are two proofs that (0, 0) is canonical, using either Int.IsCanonical.nonpos or Int.IsCanonical.nonneg. This is not an issue because by proof irrelevance these proofs must be equal.

## 12.6 Summary of Newly Introduced Lean Constructs

#### **Declaration**

noncomputable prefixes noncomputable declarations

#### **Constants**

Classical.choice function that returns an arbitrary element of a nonempty type

Classical.choose function that returns a witness given a proof of an  $\exists$ 

Quotient function that creates a quotient type of a given setoid instance

Quotient.lift function that lifts a unary function to a quotient type function that lifts a binary function to a quotient type setoid type class for a type with an equivalence relation on it

Sort u universe at level u

Subtype.mk function that constructs a subtype value

Subtype.property function that extracts the property from a subtype value function that extracts the base value from a subtype value

Notations

 $\{x : \alpha // P[x]\}$  subtype of all x in  $\alpha$  fulfilling P[x]

pprox equivalence relation on a setoid (used for quotienting)

Prop abbreviation for Sort o
Type u abbreviation for Sort (u + 1)

#### **Theorems**

Classical.axiomOfChoice

traditional axiom of choice

Classical.choose\_spec

characteristic property of Classical.choose

Quotient.exact equality on quotient type implies  $\approx$  on base type

Quotient.inductionOn

induction principle on a quotient type value

Quotient.sound  $\approx$  on base type implies equality on quotient type subtype.eq equality on base type implies equality of subtype

## **Chapter 13**

## **Basic Mathematical Structures**

In this chapter, we introduce definitions and proofs about basic mathematical structures such as groups, fields, and linear orders.

## 13.1 Type Classes over a Single Binary Operator

In mathematics, a *group* is a set G with a binary operator  $\cdot : G \times G \to G$  fulfilling the following properties, called group axioms:

- associativity: for all  $a, b, c \in G$ , we have  $(a \cdot b) \cdot c = a \cdot (b \cdot c)$ ;
- identity element: there exists an element  $e \in G$  such that for all  $a \in G$ , we have  $e \cdot a = a$ ;
- inverse element: for each  $a \in G$ , there exists an inverse element denoted  $a^{-1}$  such that  $a^{-1} \cdot a = e$ .

In Lean, a type class for groups could be defined as follows:

```
class Group (\alpha : Type) where mul : \alpha \to \alpha \to \alpha one : \alpha inv : \alpha \to \alpha mul_assoc : \forall a b c, mul (mul a b) c = mul a (mul b c) one_mul : \forall a, mul one a = a mul_left_inv : \forall a, mul (inv a) a = one
```

However, this is not the official definition. Groups are part of a larger hierarchy of algebraic structures.

Group operations can be written multiplicatively (with operator \*, identity element 1, and inverse element  $a^{-1}$ ) or additively (with operator \*, identity element 0, and inverse element -a). This is why Lean offers two type classes for groups: the multiplicative Group and the additive AddGroup. They are essentially the same but use different names for their constants and properties.

Any type that satisfies the group axioms can be registered as a Group or Add-Group. To illustrate this, we will define the type Int2 of integers modulo 2, also known as  $\mathbb{Z}/2\mathbb{Z}$  or  $\mathbb{Z}_2$ , and register it as an AddGroup. The type Int2 has two elements:

```
inductive Int2 : Type
  | zero
```

l one

Addition is defined as follows:

```
def Int2.add : Int2 → Int2 → Int2
  | Int2.zero, a => a
  | Int2.one, Int2.zero => Int2.one
  | Int2.one, Int2.one => Int2.zero
```

To instantiate AddGroup, we need to provide the following constants and properties:

```
\begin{array}{c} \operatorname{add}: \, \alpha \to \alpha \to \alpha \\ \operatorname{zero}: \, \alpha \\ \operatorname{neg}: \, \alpha \to \alpha \\ \operatorname{add\_assoc}: \, \forall \operatorname{a} \operatorname{b} \operatorname{c}, \operatorname{add} \left(\operatorname{add} \operatorname{a} \operatorname{b}\right) \operatorname{c} = \operatorname{add} \operatorname{a} \left(\operatorname{add} \operatorname{b} \operatorname{c}\right) \\ \operatorname{zero\_add}: \, \forall \operatorname{a}, \operatorname{add} \operatorname{zero} \operatorname{a} = \operatorname{a} \\ \operatorname{add\_zero}: \, \forall \operatorname{a}, \operatorname{add} \operatorname{a} \operatorname{zero} = \operatorname{a} \\ \operatorname{add\_left\_neg}: \, \forall \operatorname{a}, \operatorname{add} \left(\operatorname{neg} \operatorname{a}\right) \operatorname{a} = \operatorname{zero} \end{array}
```

The constants AddGroup.add, AddGroup.zero, and AddGroup.neg correspond to the binary operator, the identity element, and the inverse. The properties Add-Group.add\_assoc, AddGroup.zero\_add, and AddGroup.add\_left\_neg correspond to the three group axioms. For technical reasons, we must also prove the redundant property AddGroup.add\_zero.

The type Int2 can be registered as a group as follows:

```
instance Int2.AddGroup : AddGroup Int2 :=
 { add
                := Int2.add
   zero
                := Int2.zero
   neg
                 := fun a \mapsto a
   add_assoc :=
     by
        intro a b c
        cases a <;>
          cases b <;>
          cases c <;>
          rfl
   zero_add
                :=
     by
        intro a
        cases a <;>
          rfl
   add_zero
     by
        intro a
        cases a <;>
          rfl
   add_left_neg :=
     by
        intro a
        cases a <;>
```

```
rfl }
```

Thanks to this instance, we can now write o, +, and -:

```
#reduce Int2.one + 0 - 0 - Int2.one
```

The algebraic hierarchy contains further type classes with one binary operator. The main ones are listed below:

Type class	Properties	Examples
Semigroup	associativity of *	$\mathbb{R}$ , $\mathbb{Q}$ , $\mathbb{Z}$ , $\mathbb{N}$
Monoid	Semigroup with unit 1	$\mathbb{R}$ , $\mathbb{Q}$ , $\mathbb{Z}$ , $\mathbb{N}$
LeftCancelSemigroup	Semigroup with $c * a = c * b \rightarrow a = b$	
RightCancelSemigroup	Semigroup with $a * c = b * c \rightarrow a = b$	
Group	Monoid with inverse $^{-1}$	

For most of these structures, commutative versions (where  $a \cdot b = b \cdot a$  for all elements a, b) are available: CommSemigroup, CommMonoid, CommGroup. They also all have additive counterparts prefixed with Add:

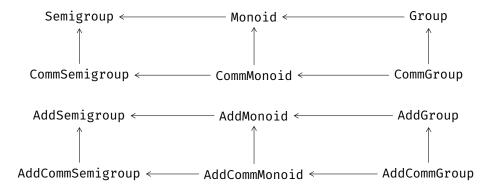
Type class	Properties	Examples
AddSemigroup	associativity of +	$\mathbb{R}$ , $\mathbb{Q}$ , $\mathbb{Z}$ , $\mathbb{N}$
AddMonoid	AddSemigroup with unit o	$\mathbb{R}$ , $\mathbb{Q}$ , $\mathbb{Z}$ , $\mathbb{N}$
AddLeftCancelSemigroup	AddSemigroup with	$\mathbb{R}$ , $\mathbb{Q}$ , $\mathbb{Z}$ , $\mathbb{N}$
	$c + a = c + b \rightarrow a = b$	
AddRightCancelSemigroup	AddSemigroup <b>with</b>	$\mathbb{R}$ , $\mathbb{Q}$ , $\mathbb{Z}$ , $\mathbb{N}$
	$a + c = b + c \rightarrow a = b$	
AddGroup	AddMonoid with inverse -	$\mathbb{R}$ , $\mathbb{Q}$ , $\mathbb{Z}$

Although the additive type classes are merely copies of their multiplicative counterparts, they are crucial when constructing algebraic structures with more than one binary operator such as rings and fields. To avoid duplicating all theorems and definitions based on the multiplicative type classes, the copying process is automated via metaprograms.

An example instance of AddMonoid is the type List  $\alpha$  with the empty list [] as zero and the append operator ++ as addition:

We could continue and register List  $\alpha$  with [] and ++ as an AddLeftCancelSemigroup and an AddRightCancelSemigroup.

The graph below illustrates the relationships between some of these type classes. An arrow from X to Y means "X inherits all constants and properties from Y."



## 13.2 Type Classes over Two Binary Operators

The additive and multiplicative structures are amalgamated to form more complex structures over two binary operators. One of these is field. A *field F* is defined by the following properties:

- F forms a commutative group under an operator +, called addition, with identity element o.
- $F\setminus\{0\}$  forms a commutative group under an operator \*, called multiplication.
- Multiplication distributes over addition—i.e., a \* (b + c) = a \* b + a \* c for all  $a, b, c \in F$ .

By running #print Field, we can display all constants and properties required by Field. Again, the type class includes some redundant properties due to its construction.

We will now show that Int2 is a field by instantiating the Field type class with it. First, we must define multiplication on Int2:

```
def Int2.mul : Int2 \rightarrow Int2 \rightarrow Int2
  | Int2.one, a => a
  | Int2.zero, _ => Int2.zero
```

To declare Int2 as a field, we can reuse the instance Int2.AddGroup that we defined above using the syntax Int2.AddGroup with. We can prove the remaining properties as follows:

```
instance Int2.Field : Field Int2 :=
 { Int2.AddGroup with
                 := Int2.one
   one
   mul
                 := Int2.mul
   inv
                 := fun a \mapsto a
   add_comm
                 :=
     by
       intro a b
       cases a <;>
         cases b <;>
         rfl
   exists_pair_ne :=
       apply Exists.intro Int2.zero
```

```
apply Exists.intro Int2.one
   simp
zero_mul
        :=
 by
   intro a
   rfl
mul_zero :=
 by
   intro a
   cases a <;>
     rfl
one_mul
         :=
 by
   intro a
   rfl
mul_one :=
 by
   intro a
   cases a <;>
    rfl
mul_inv_cancel :=
 by
   intro a h
   cases a
   { apply False.elim
    apply h
    rfl }
   { rfl }
inv_zero := by rfl
mul_assoc :=
 by
   intro a b c
   cases a <;>
    cases b <;>
     cases c <;>
     rfl
mul_comm :=
 by
   intro a b
   cases a <;>
     cases b <;>
     rfl
left_distrib :=
 by
   intro a b c
   cases a <;>
     cases b <;>
     rfl
right_distrib :=
```

```
by
  intro a b c
  cases a <;>
    cases b <;>
    cases c <;>
    rfl }
```

With this declaration in place, we can now use the notations 1, \*, /, and more:

```
#reduce (1 : Int2) * 0 / (0 - 1)
```

This command prints Int2.zero. The type annotation: Int2 is necessary here to tell Lean that we want to calculate in Int2 and not in  $\mathbb{N}$ , the default. We can even use arbitrary numerals in Int2. For example, the numeral 3 is interpreted as 1 + 1 + 1, which is the same as 1 in Int2:

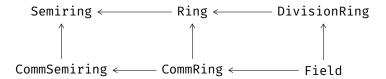
```
#reduce (3 : Int2)
```

This command prints Int2.one.

Besides Field, there are many more type classes for structures with two binary operators. These are the main ones:

Type class	Properties	Examples
Semiring	Monoid and AddCommMonoid with distributivity	$\mathbb{R}$ , $\mathbb{Q}$ , $\mathbb{Z}$ , $\mathbb{N}$
CommSemiring	Semiring with commutativity of *	$\mathbb{R}$ , $\mathbb{Q}$ , $\mathbb{Z}$ , $\mathbb{N}$
Ring	Monoid and AddCommGroup with distributivity	$\mathbb{R}$ , $\mathbb{Q}$ , $\mathbb{Z}$
CommRing	Ring with commutativity of *	$\mathbb{R}$ , $\mathbb{Q}$ , $\mathbb{Z}$
DivisionRing	Ring with multiplicative inverse	$\mathbb{R}$ , $\mathbb{Q}$
Field	DivisionRing with commutativity of *	$\mathbb{R}$ , $\mathbb{Q}$

The graph below illustrates the relationships between these type classes.



The hierarchy between Ring and Field is more complex than depicted and includes type classes for domains, integral domains, and Euclidean rings.

The Field type class requires the property  $\forall a$ , a / o = o. This is simply a convention to make division a total function. Mathematicians would regard division as a partial function. There is no harm in totalizing a partial function in this way.

Once we have instantiated the type classes with types of interest, we can use the ring tactic to normalize terms containing operators on those types. For example:

```
theorem ring_example (a b : Int2) :
  (a + b) ^ 3 = a ^ 3 + 3 * a ^ 2 * b + 3 * a * b ^ 2 + b ^ 3 :=
  by ring
```

This tactic is available for any type that is declared as a field or more generally as a commutative semiring.

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### 13.3 Coercions

When combining numbers from  $\mathbb{N}$ ,  $\mathbb{Z}$ ,  $\mathbb{Q}$ , and  $\mathbb{R}$  in the same theorems, we may want to cast from one type to another. For example, given a natural number, we may need to convert it to an integer. Consider the following theorem and notice the type of the arguments of multiplication:

```
theorem neg_mul_neg_Nat (n : \mathbb{N}) (z : \mathbb{Z}) :

(-z) * (-n) = z * n :=

by simp
```

Surprisingly, this statement does not lead to an error, although negation - n is not defined on  $n : \mathbb{N}$ , and multiplication of  $z : \mathbb{Z}$  and  $n : \mathbb{N}$  is not defined.

The diagnostic command #check neg\_mul\_neg\_Nat tells us what happened:

```
neg_mul_neg_Nat : \forall (n : \mathbb{N}) (z : \mathbb{Z}), -z * -\uparrow n = z * \uparrow n
```

Lean has a mechanism to introduce coercions, denoted by  $\uparrow$  or coe, when necessary. This coercion operator can be set up to provide implicit conversions between arbitrary types. Many coercions are already in place, including these:

- coe :  $\mathbb{N} \to \alpha$  casts  $\mathbb{N}$  to another semiring  $\alpha$ ;
- coe :  $\mathbb{Z} \to \alpha$  casts  $\mathbb{Z}$  to another ring  $\alpha$ ;
- coe :  $\mathbb{Q} \to \alpha$  casts  $\mathbb{Q}$  to another division ring  $\alpha$ .

We can provide type annotations to document our intentions or help Lean figure out where to place the coercions, as in the following example:

```
theorem neg_Nat_mul_neg(n : \mathbb{N})(z : \mathbb{Z}):
(-n : \mathbb{Z}) * (-z) = n * z :=
by simp
```

In proofs involving coercions, the norm\_cast tactic can be convenient. It helps with goals such as  $\vdash$  m n :  $\mathbb{N}$ , h :  $\uparrow$ m =  $\uparrow$ n  $\vdash$  m = n in

```
theorem Eq_coe_int_imp_Eq_Nat (m n : N)
    (h : (m : Z) = (n : Z)) :
    m = n :=
    by norm_cast at h
```

Similarly, it helps with the goal  $\vdash$  m n :  $\mathbb{N} \vdash \uparrow$ m +  $\uparrow$ n =  $\uparrow$ (m + n) in

```
theorem Nat_coe_Int_add_eq_add_Nat_coe_Int (m n : \mathbb{N}) : (m : \mathbb{Z}) + (n : \mathbb{Z}) = ((m + n : \mathbb{N}) : \mathbb{Z}) := by norm_cast
```

The norm\_cast tactic relies on theorems such as the following:

```
Nat.cast_add : \forall a \ b : \mathbb{N}, \uparrow(a + b) = \uparrow a + \uparrow b

Int.cast_add : \forall a \ b : \mathbb{Z}, \uparrow(a + b) = \uparrow a + \uparrow b

Rat.cast_add : \forall a \ b : \mathbb{Q}, \uparrow(a + b) = \uparrow a + \uparrow b
```

### 13.4 Normalization Tactics

The algebraic tactic ring and the coercion tactic norm\_cast work by normalization: They rewrite expressions in the hope that they become syntactically equal, at

which point equality is trivial to prove. Like rw and simp, they produce a subgoal when they make some progress but do not fully prove the goal.

The optional position argument is as for the rewriting tactics (Section 3.5).

#### ring

```
ring [at position]
```

The ring tactic proves equalities over commutative rings and semirings (such as  $\mathbb{N}$ ,  $\mathbb{Z}$ ,  $\mathbb{Q}$ , and  $\mathbb{R}$ ) by normalizing expressions and syntactically comparing the result.

#### norm\_cast

```
norm_cast [at position]
```

The norm\_cast tactic moves coercions towards the inside of expressions, as a form of simplification.

### 13.5 Lists, Multisets, and Finite Sets

We have seen many examples of how lists can be used in previous chapters. But when making a new definition or stating a new theorem, we should also reflect on alternatives such as multisets and finite sets.

Consider the following definition, based on the binary trees we introduced in Section 5.8:

This function returns a list of all the elements occurring in the tree. It traverses the tree depth first, from left to right. But for some applications, we might not care about the order of the elements.

This is where multisets come into play. For multisets, we have  $\{3, 2, 1, 2\} = \{1, 2, 2, 3\}$ , whereas the two lists [3, 2, 1, 2] and [1, 2, 2, 3] are different. Multisets are defined as the quotient type over lists up to reordering. We can redo the above definition using multisets as follows:

Using this definition, we can prove that Multiset.elems t = Multiset.elems (mirror t), whereas List.elems t = List.elems (mirror t) does not hold in general.

For some applications, we might want to go a step further and ignore not only the order but also how often each element occurs in the tree, distinguishing only between occurrence and nonoccurrence. This is where finite sets, or *finsets*, come into play. On finsets, we have  $\{3, 2, 1, 2\} = \{1, 2, 3\}$ . Finsets are defined as the subtype of multisets that do not contain any repeated elements. (Another

possible definition would have been as the subtype of sets that are finite.) We can redo the definition above using finsets as follows:

```
def Finset.elems : BinTree \mathbb{N} → Finset \mathbb{N} | BinTree.empty => \emptyset | BinTree.node a l r => {a} \cup Finset.elems l \cup Finset.elems r
```

For lists and multisets, Lean offers sum and product operators to add or multiply all of the elements. Below, the first two commands print 9, and the last two print 24:

```
#eval List.sum [2, 3, 4]
#eval Multiset.sum (\{2, 3, 4\}: Multiset \mathbb{N})
#eval List.prod [2, 3, 4]
#eval Multiset.prod (\{2, 3, 4\}: Multiset \mathbb{N})
```

These operators require the type of the elements to be declared as an instance of AddMonoid for sum or of Monoid for product. The Multiset and Finset versions also require an instance declaration for AddCommMonoid or CommMonoid because the result cannot depend on the order of adding or multiplying the elements.

## **13.6 Order Type Classes**

Many of the structures introduced above can be ordered. For example, the familiar order on the natural numbers can be defined as

```
inductive Nat.le : \mathbb{N} \to \mathbb{N} \to \mathsf{Prop}
| refl : \forall a : \mathbb{N}, Nat.le a a
| step : \forall a b : \mathbb{N}, Nat.le a b \to Nat.le a (b + 1)
```

This is an example of a linear order. A linear order (or total order) is a binary relation  $\leq$  such that for all a, b, and c, the following properties hold:

- reflexivity:  $a \le a$ ;
- transitivity: if a < b and b < c, then a < c;
- antisymmetry: if  $a \le b$  and  $b \le a$ , then a = b;
- totality:  $a \le b$  or  $b \le a$ .

If a relation has the first three properties, it is a *partial order*. An example is the subset relation  $\subseteq$  on sets, finite sets, or multisets. If a relation has the first two properties, it is a *preorder*. An example is comparing lists by their lengths.

In Lean, there are type classes for the different kinds of orders: LinearOrder, PartialOrder, and Preorder. The Preorder class has one constant and two properties:

```
\begin{array}{c} \text{le}: \alpha \rightarrow \alpha \rightarrow \text{Prop} \\ \text{le\_refl}: \forall \text{a}: \alpha \text{, le a a} \\ \text{le\_trans}: \forall \text{a b c}: \alpha \text{, le a b} \rightarrow \text{le b c} \rightarrow \text{le a c} \end{array}
```

The PartialOrder class has the additional property

```
le antisymm : \forall a b : \alpha, le a b \rightarrow le b a \rightarrow a = b
```

and LinearOrder has the additional property

```
le_total : \foralla b : \alpha, le a b \vee le b a
```

We can declare the preorder on List  $\alpha$  that compares lists by their lengths as follows:

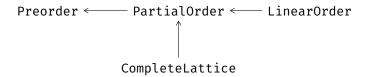
```
instance List.length.Preorder \{\alpha : \mathsf{Type}\} : \mathsf{Preorder}(\mathsf{List}\ \alpha) :=
       fun xs ys \mapsto List.length xs \leq List.length ys
       fun xs ys → List.length xs < List.length ys</pre>
    le refl :=
      by
         intro xs
         apply Nat.le_refl
    le trans :=
      by
         intro xs ys zs
         exact Nat.le_trans
    lt_iff_le_not_le :=
       by
         intro a b
         simp
         intro hlt
         linarith }
```

This type class instance gives access to the infix syntax  $\leq$  and to the corresponding relations  $\geq$ , <, and >:

```
theorem list.length.Preorder_example :
  [1] > [] :=
  by decide
```

The proof features a new tactic: decide. It relies on type class inference to prove trivial decidable goals.

Complete lattices, which we discussed in Chapter 11, are formalized as another type class, CompleteLattice, which extends PartialOrder. In the following diagram, an arrow from X to Y means "X inherits all constants and properties from Y."



Finally, Lean provides type classes that combine orders and algebraic structures: OrderedCancelCommMonoid, OrderedCommGroup, OrderedSemiring, Linear-OrderedSemiring, LinearOrderedCommRing, LinearOrderedField. All these mathematical structures relate  $\leq$  and < with the constants 0, 1, +, and  $\star$  by monotonicity rules (e.g., a  $\leq$  b  $\rightarrow$  c  $\leq$  d  $\rightarrow$  a + c  $\leq$  b + d) and cancellation rules (e.g., c + a < c + b  $\rightarrow$  a < b).

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## 13.7 Decision Tactic

#### decide

The decide tactic can be used to prove true decidable goals. Decidability is determined by checking membership in the Decidable type class. Unlike rfl, decide is not limited to proving equalities.

## **13.8 Summary of Newly Introduced Lean Constructs**

#### **Notation**

↑ coercion operator coe

#### **Tactics**

decide proves decidable truths (e.g., a true executable expression)

ring normalizes ring expressions

## **Chapter 14**

## **Rational and Real Numbers**

We have seen how the natural numbers  $\mathbb N$  can be defined as an inductive type and how the integers  $\mathbb Z$  can be defined as a quotient over  $\mathbb N \times \mathbb N$ . In this chapter, we review the construction of the rational numbers  $\mathbb Q$  and the real numbers  $\mathbb R$ . The tools used for these constructions are inductive types, subtypes, and quotients.

The following procedure can be used to construct types that enjoy specific properties:

- 1. Create a new type that is large enough to represent all elements, but not necessarily in a unique manner.
- 2. Take the quotient of this representation, equating elements as needed.
- 3. Define operators on the quotient type by lifting functions from the base type, and prove that they are compatible with the quotient relation.

We used this approach to construct the type  $\mathbb Z$  in Section 12.5.1. It can also be used for  $\mathbb Q$  and  $\mathbb R$ .

### **14.1 Rational Numbers**

A rational number is a number that can be expressed as a fraction n/d of integers n and d, where  $d \neq 0$ :

The number n is called the numerator, and the number d is called the denominator. The representation of a rational number as a fraction is not unique. For example, the rationals 1/2, 2/4, and -1/-2 are all equal. This representation as a fraction will serve as the base type of which we will take the quotient.

Two fractions  $n_1/d_1$  and  $n_2/d_2$  represent the same rational number if the ratio between numerator and denominator are the same:  $n_1*d_2=n_2*d_1$ . To construct the quotient over the type Fraction with respect to this relation, we show that the relation is an equivalence relation. This is achieved by declaring Fraction an instance of the Setoid type class:

```
namespace Fraction
```

```
instance Setoid : Setoid Fraction :=
  { r :=
      fun a b : Fraction \mapsto num a * denom b = num b * denom a
    iseqv :=
      { refl := by aesop
        symm := by aesop
        trans :=
          by
            intro a b c heq_ab heq_bc
            apply Int.eq_of_mul_eq_mul_right (denom_Neq_zero b)
            calc
               num a * denom c * denom b
               = num a * denom b * denom c :=
                 by ac_rfl
               _ = num b * denom a * denom c :=
                 by rw [heq_ab]
               = num b * denom c * denom a :=
                 by ac_rfl
               _ = num c * denom b * denom a :=
                 by rw [heq_bc]
               _ = num c * denom a * denom b :=
                 by ac rfl
             } }
theorem Setoid_Iff (a b : Fraction) :
  a \approx b \leftrightarrow num \ a * denom \ b = num \ b * denom \ a :=
  by rfl
end Fraction
```

Then we can define the type of rationals as the quotient over this setoid:

```
def Rat : Type :=
  Quotient Fraction.Setoid
```

To define zero, one, addition, multiplication, and other operations, we first define them on the Fraction type. To add two fractions, we convert them to a common denominator and add the numerators. The easiest common denominator to use is simply the product of the two denominators:

We register these operations as instances of the syntactic type classes such as Add to be able to use convenient notation such as + on Fraction. Similarly, we define zero as 0 := 0/1, one as 1 := 1/1, and multiplication as the pairwise multiplication of numerators and denominators.

To lift these operations to the type of rational numbers,  ${\tt Rat},$  we must prove them compatible with  $\approx\!\!:$ 

```
namespace Fraction
a[simp] theorem add_num (a b : Fraction) :
  num(a + b) = num a * denom b + num b * denom a :=
  by rfl
a[simp] theorem add_denom (a b : Fraction) :
  denom (a + b) = denom a * denom b :=
  by rfl
theorem Setoid_add {a a' b b' : Fraction} (ha : a \approx a')
    (hb : b \approx b') :
  a + b \approx a' + b' :=
  by
    simp [Setoid_Iff, add_denom, add_num] at *
      (num a * denom b + num b * denom a)
          * (denom a' * denom b')
      = num a * denom a' * denom b * denom b'
        + num b * denom b' * denom a * denom a' :=
        by
          simp [add_mul, mul_add]
          ac_rfl
      _ = num a' * denom a * denom b * denom b'
            + num b' * denom b * denom a * denom a' :=
        by simp [*]
      _ = (num a' * denom b' + num b' * denom a')
            * (denom a * denom b) :=
        by
          simp [add_mul, mul_add]
          ac_rfl
end Fraction
```

Then we can use Quotient.lift(2) to define the Rat operations, and we can instantiate the relevant syntactic type classes, such as

```
namespace Rat
instance Add : Add Rat :=
  { add := Quotient.lift₂ (fun a b : Fraction → mk (a + b))
        (by
            intro a b a' b' ha hb
            apply Quotient.sound
            exact Fraction.Setoid_add ha hb) }
end Rat
```

From here, we can proceed and prove all the properties needed to make Rat an instance of Field.

Alternative Definitions of the Rational Numbers In mathlib, another approach is used to define the rationals. The type Rat is defined as a subtype of Fraction, with the requirement that the denominator is positive and that the numerator and the denominator are coprime (i.e., they have no common divisors except 1 and -1):

This is an instance of the general strategy described in Section 12.5.3. With this approach, no quotient is required, computation is more efficient, and more properties are syntactic equalities up to computation. A drawback is that function definitions become more complicated due to the need to normalize fractions.

#### 14.2 Real Numbers

Some sequences of rational numbers seem to converge because the numbers in the sequence get closer and closer to each other and yet do not converge to a rational number. The sequence

```
a_0 = 1
a_1 = 1.4
a_2 = 1.41
a_3 = 1.414
a_4 = 1.4142
:
```

where  $a_n$  is the largest number with n digits after the decimal point such that  $a_n^2 < 2$ , is such a sequence. It seems to converge because each  $a_n$  is at most  $10^{-n}$  away from any of the following numbers, but the limit is  $\sqrt{2} \notin \mathbb{Q}$ . In that sense, the rational numbers are incomplete, and the reals are their *completion*. To construct the reals, we need to fill in the gaps that are revealed by these sequences that seem to converge but do not.

Cauchy sequences capture the notion of a sequence that seems to converge. A sequence  $a_0, a_1, \ldots$  is *Cauchy* if for any  $\varepsilon > 0$ , there exists an  $N \in \mathbb{N}$  such that for all  $m \geq N$ , we have  $|a_N - a_m| < \varepsilon$ . In other words, no matter how small we choose  $\varepsilon$ , we can always find a point in the sequence from which all following numbers deviate by less than  $\varepsilon$ .

We formalize sequences of rational numbers as functions  $f:\mathbb{N}\to\mathbb{Q}$  and denote the absolute value  $|\ |$  by abs. This yields the following Lean definition of Cauchy sequences:

```
def IsCauchySeq (f : \mathbb{N} \to \mathbb{Q}) : Prop := \forall \varepsilon > 0, \exists N, \forall m \geq N, abs (f N - f m) < \varepsilon
```

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Not every sequence is a Cauchy sequence:

```
theorem id_Not_CauchySeq :
  ¬ IsCauchySeq (fun n : N → (n : Q)) :=
by
  rw [IsCauchySeq]
  intro h
  cases h 1 zero_lt_one with
  | intro i hi =>
   have hi_succi :=
    hi (i + 1) (by simp)
  simp [←sub_sub] at hi_succi
```

We define a type of Cauchy sequences as a subtype:

```
\begin{array}{lll} \text{def CauchySeq} & : & \text{Type} & := \\ & \{f: \ \mathbb{N} \ \rightarrow \ \mathbb{Q} \ \ // \ \text{IsCauchySeq} \ f \} \end{array}
```

It will be convenient to have an auxiliary function that extracts the actual sequence from a CauchySeq:

```
\begin{array}{l} \operatorname{\mathsf{def}} \ \operatorname{\mathsf{seq0f}} \ (\mathsf{f} : \mathsf{CauchySeq}) : \mathbb{N} \to \mathbb{Q} := \\ \operatorname{\mathsf{Subtype.val}} \ \mathsf{f} \end{array}
```

The basic idea of the construction is to represent the real numbers by Cauchy sequences. Each Cauchy sequence represents the real number that is its limit; for example, the sequence  $a_n = 1/n$  represents the real number o, and the sequence 1, 1.4, 1.41, ... represents the real number  $\sqrt{2}$ .

Two different Cauchy sequences can represent the same real number; for example, the sequence  $a_n = 1/n$  and the constant sequence  $b_n = 0$  both represent o. Therefore, we need to take the quotient over sequences representing the same real number. Two sequences represent the same real number when their difference converges to zero:

```
namespace CauchySeq
instance Setoid : Setoid CauchySeq :=
{ r :=
     fun f g : CauchySeq \mapsto
        \forall \varepsilon > 0, \exists N, \forall m \geq N, abs (seqOf f m - seqOf g m) < \varepsilon
  iseqv :=
     { refl :=
           by
              intro f \varepsilon h\varepsilon
              apply Exists.intro o
              aesop
        symm :=
           by
              intro f g hfg \varepsilon h\varepsilon
              cases hfg \varepsilon h\varepsilon with
              | intro N hN =>
                apply Exists.intro N
                 intro m hm
```

```
rw [abs_sub_comm]
                apply hN m hm
        trans :=
          bν
             intro f g h hfg hgh \varepsilon h\varepsilon
             cases hfg (\varepsilon / 2) (half_pos h\varepsilon) with
             | intro N_1 hN_1 = >
                cases hgh (\varepsilon / 2) (half_pos h\varepsilon) with
                | intro N<sub>2</sub> hN<sub>2</sub> =>
                  apply Exists.intro (max N_1 N_2)
                  intro m hm
                  calc
                     abs (seqOf f m - seqOf h m)
                     \leq abs (seqOf f m - seqOf g m)
                        + abs (seqOf g m - seqOf h m) :=
                        by apply abs_sub_le
                   _ < ε / 2 + ε / 2 :=
                     add_lt_add (hN<sub>1</sub> m (le_of_max_le_left hm))
                        (hN<sub>2</sub> m (le_of_max_le_right hm))
                  _ = ε :=
                     by simp } }
theorem Setoid_iff (f g : CauchySeq) :
  f \approx g \leftrightarrow
  \forall \varepsilon > 0, \exists N, \forall m \geq N, abs (seqOf f m - seqOf g m) < \varepsilon :=
  by rfl
end CauchySeq
```

Using this Setoid instance, we can now define the real numbers:

```
def Real : Type :=
  Quotient CauchySeq.Setoid
```

As with the rational numbers, we need to define zero, one, addition, multiplication, and other operators. We define them on CauchySeq first and lift them to Real afterwards. For the constants 0 and 1, we can define them simply as constant sequences. Any constant sequence is a Cauchy sequence:

We can declare Real instances of the syntactic type classes Zero and One:

```
instance Zero : Zero Real :=
  { zero := [CauchySeq.const o] }
```

```
instance One : One Real :=
  { one := [CauchySeq.const 1] }
```

Defining addition of real numbers requires a little more effort. We define addition on Cauchy sequences by adding the elements of the sequence pairwise:

This definition requires a proof that the result is a Cauchy sequence, given that f and g are Cauchy sequences. It is omitted, hence the sorry.

Next, we need to show that this addition is compatible with  $\approx$ :

```
theorem Setoid_add {f f' g g' : CauchySeq} (hf : f \approx f')
    (hg : g \approx g') :
  f + g \approx f' + g' :=
  by
    intro \varepsilon_0 h\varepsilon_0
    simp [Setoid_iff]
    cases hf (\varepsilon_0 / 2) (half_pos h\varepsilon_0) with
    | intro Nf hNf =>
       cases hg (\epsilon_{\rm 0} / 2) (half_pos h\epsilon_{\rm 0}) with
       | intro Ng hNg =>
         apply Exists.intro (max Nf Ng)
         intro m hm
         calc
           abs (seqOf (f + g) m - seqOf (f' + g') m)
           = abs ((seqOf f m + seqOf g m)
              - (seqOf f' m + seqOf g' m)) :=
             by rfl
           _{-} = abs ((seqOf f m - seqOf f' m)
                + (seqOf g m - seqOf g' m)) :=
              by
                have arg_eq :
                  seqOf f m + seqOf g m
                     - (seqOf f' m + seqOf g' m) =
                  seqOf f m - seqOf f' m
                     + (seqOf g m - seqOf g' m) :=
                  by linarith
                rw [arg_eq]
           _{-} \leq abs (seq0f f m - seq0f f' m)
                + abs (seqOf g m - seqOf g' m) :=
              by apply abs_add
           _{-} < \epsilon_{0} / 2 + \epsilon_{0} / 2 :=
              add_lt_add (hNf m (le_of_max_le_left hm))
                (hNg m (le of max le right hm))
           _{-} = \varepsilon_{\odot} :=
              by simp
```

To prove that f + g  $\approx$  f' + g', we are given an  $\epsilon_0$  > 0 and must show that there exists a number N such that

```
\forallm, m\geqN\rightarrowabs (seqOf(f+g)m - seqOf(f'+g')m) < \epsilon_{\Theta}
```

To obtain N, we use  $f \approx f'$  and  $g \approx g'$ . The equivalence  $f \approx f'$  gives us for any  $\varepsilon > 0$  a number Nf such that abs (seqOf f m - seqOf f' m) <  $\varepsilon$  for all m  $\geq$  Nf. The fact  $g \approx g'$  gives us a number Ng with a similar property. For the calculations to work out in the end, we take the numbers Nf and Ng for  $\varepsilon := \varepsilon_0 / 2$ . Then we choose N to be the maximum of Nf and Ng, so that we get the inequalities for any m  $\geq$  N. The calc block at the end of the proof establishes that, for all m  $\geq$  N,

```
abs (seqOf (f + g) m - seqOf (f' + g') m) < \varepsilon_0
```

Having shown that addition on Cauchy sequences is compatible with  $\approx$ , we can define addition on Real:

```
namespace Real
instance Add : Add Real :=
{ add := Quotient.lift₂ (fun a b : CauchySeq → [a + b])
    (by
        intro a b a' b' ha hb
        apply Quotient.sound
        exact CauchySeq.Setoid_add ha hb) }
end Real
```

We could continue like this with multiplication and other operators. In summary, real numbers are defined as a quotient over Cauchy sequences, which are in turn defined as a subtype of  $\mathbb{N} \to \mathbb{Q}$ .

Alternative Definitions of the Real Numbers In mathlib, the construction of the real numbers is essentially as described above. Some definitions are stated in a more general fashion to allow construction of other algebraic structures, such as the *p*-adic numbers [18].

Alternatively, the real numbers can be defined using Dedekind cuts. A number  $r:\mathbb{R}$  is then represented as the set of numbers  $x:\mathbb{Q}$  such that x< r. Another alternative, which does not depend on  $\mathbb{Q}$ , is to define  $\mathbb{R}$  using binary sequences  $\mathbb{N} \to \mathsf{Bool}$ . The elements of the sequence represent the number's digits. This works particularly well if we only need the real interval [0,1].

## 14.3 Final Exhortation

We have now reached the end of this guide. You now know fundamental theory and techniques in interactive theorem proving as well as some application areas. Although we used Lean, what you have learned should translate to other systems, especially those based on simple or dependent type theory. And you ought to be able to read most scientific papers in the area.

Even if you do not choose to pursue a career in theorem proving, the authors hope you will bring proof assistants with you and use them when it makes sense,

either because of their high trustworthiness or because of their convenience for keeping track of complex proof goals.

If you continue using Lean, the natural next step would be to familiarize your-self with mathlib and its documentation. If you use Lean outside a class context, you will often find yourself looking up definitions and theorems. You will surely find the #find command useful. And most Lean users use the Lean Zulip chat.<sup>2</sup>

## **14.4 Summary of Newly Introduced Lean Constructs**

## **Diagnostic Command**

#find

looks up a definition or theorem by pattern matching

https://leanprover-community.github.io/mathlib4\_docs/Mathlib/Tactic/Find.html

<sup>2</sup>https://leanprover.zulipchat.com/

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